

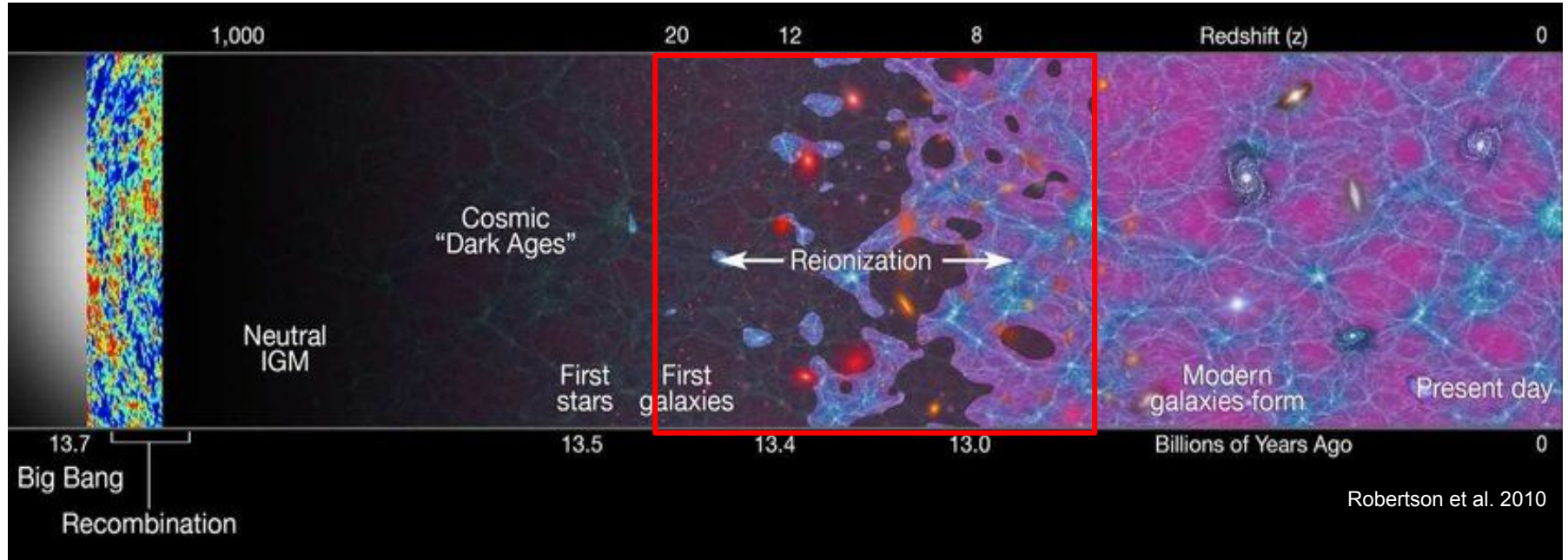
# Constraining the mean free path of ionizing photons ( $\lambda_{\text{mfp}}$ ) at $z > 5$ from the Lyman- $\alpha$ forest flux auto-correlation function

2208.09013

Molly Wolfson (UCSB)



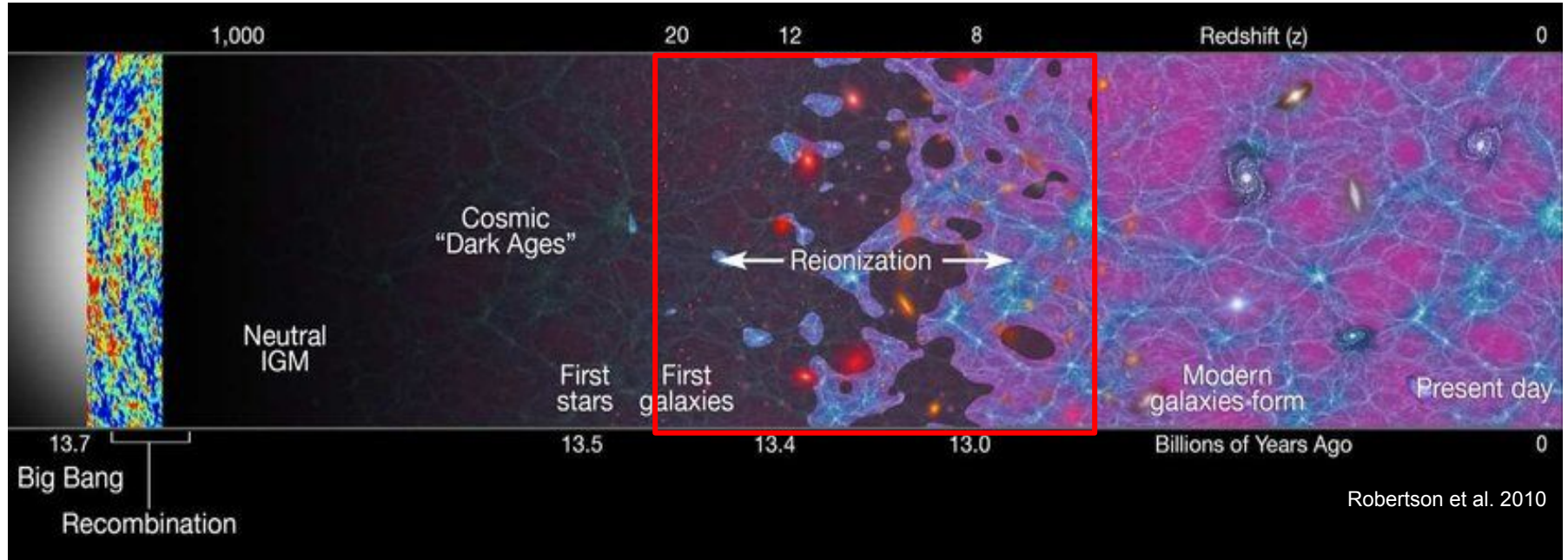
# What is reionization?



Planck gives the midpoint:  $z_{\text{re}} \sim 7.7$

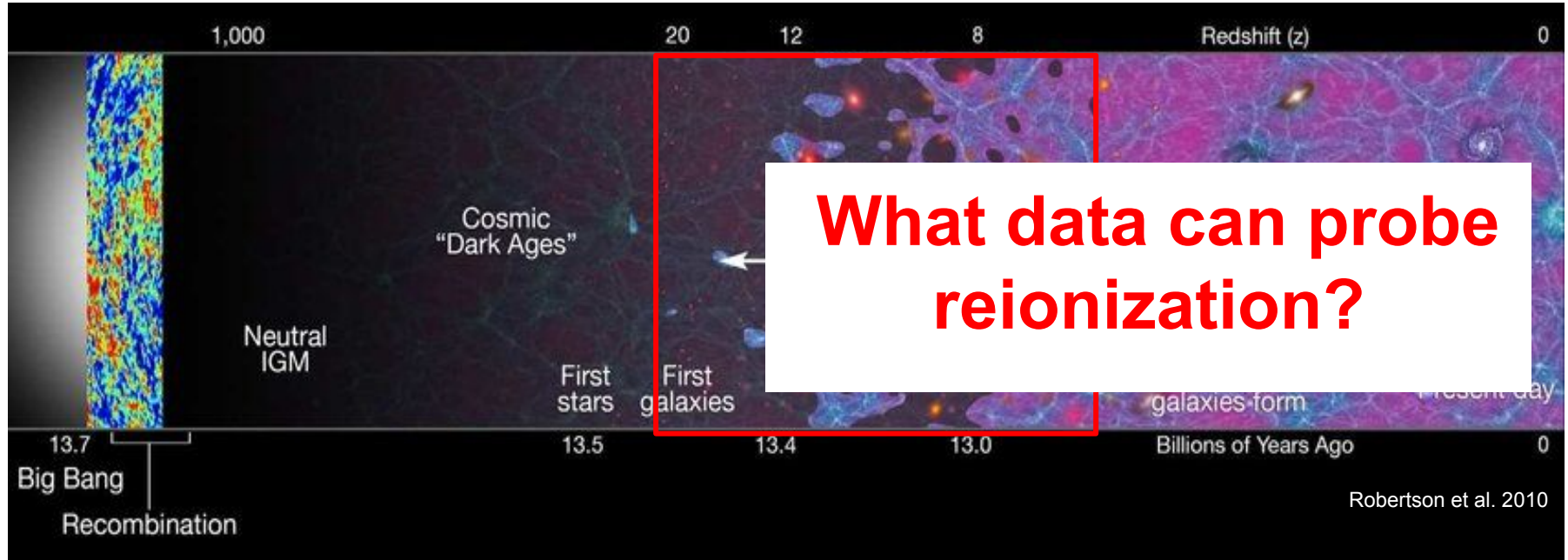
Recent measurements suggest that it is not complete until  $z < 6$

# What is reionization?



A lot still unknown (driving sources, number of photons required, impact on thermal state of IGM, and more)

# What is reionization?



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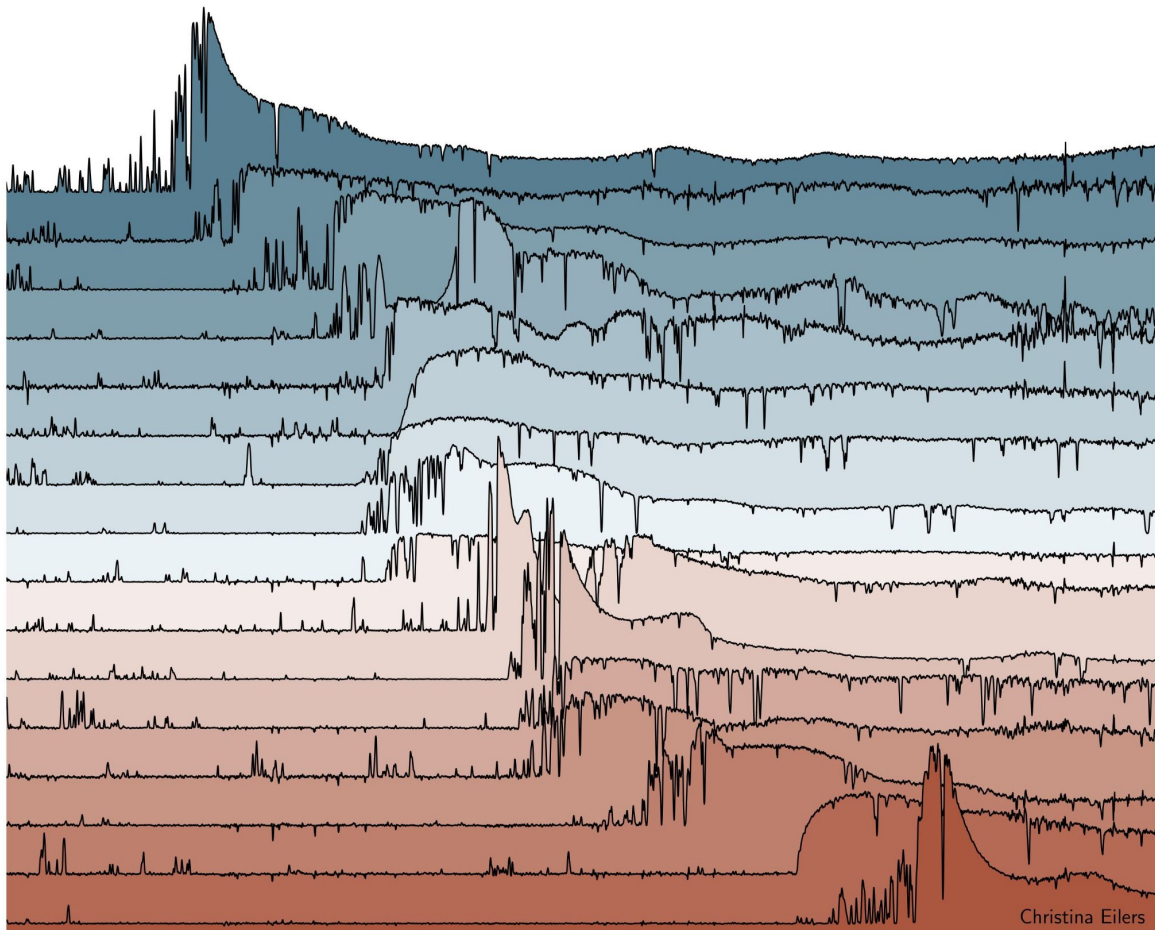


# High redshift quasars

# XQR-30 Collaboration

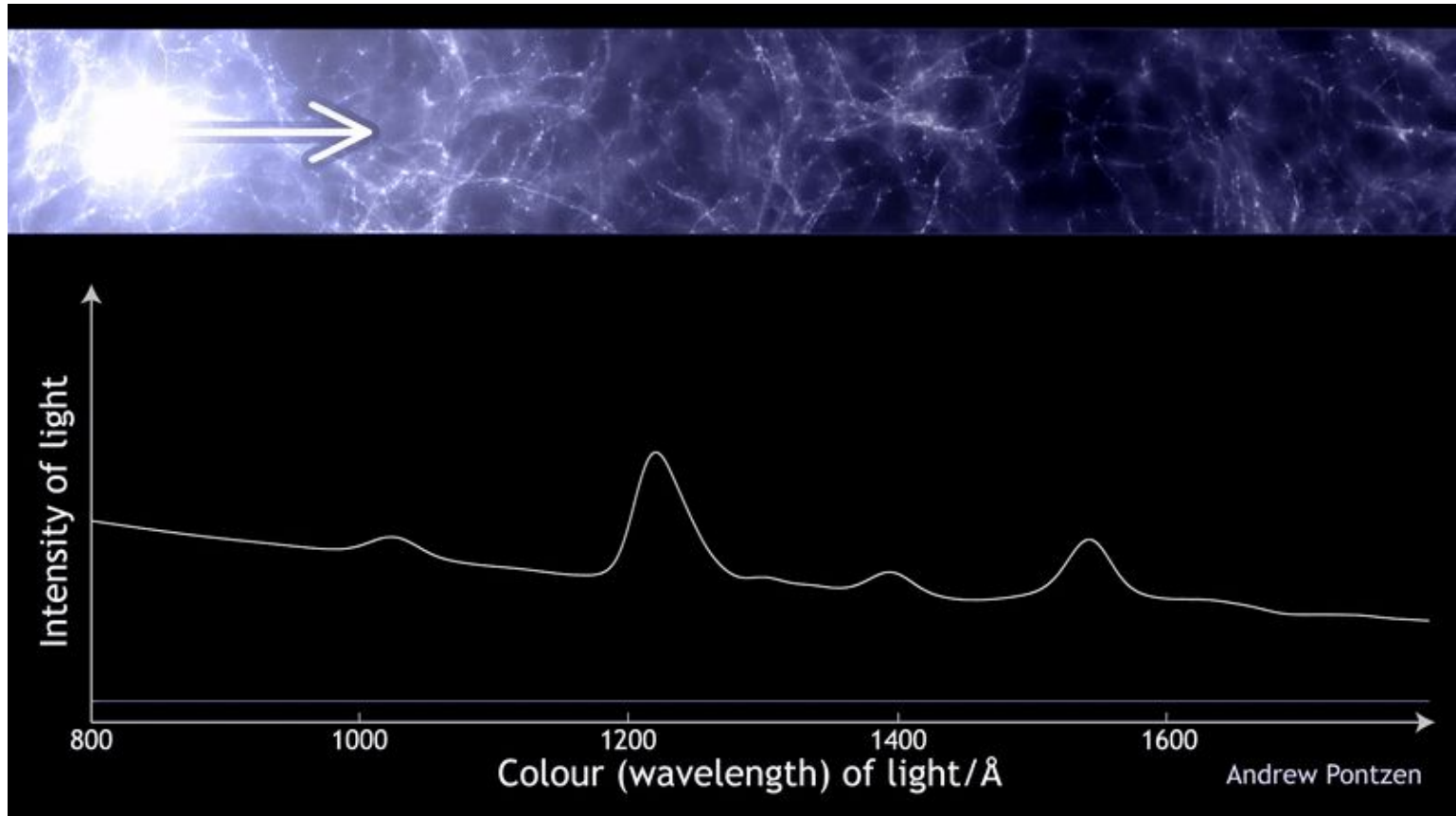
XQR-30 data ([xqr30.inaf.it](http://xqr30.inaf.it)):

- Uses VLT/X-Shooter (R ~ 8800 in the visible)
- 30 new observations of some of the brightest  $z > 5.8$  quasars observed
- Supplemented with 12 archival observations

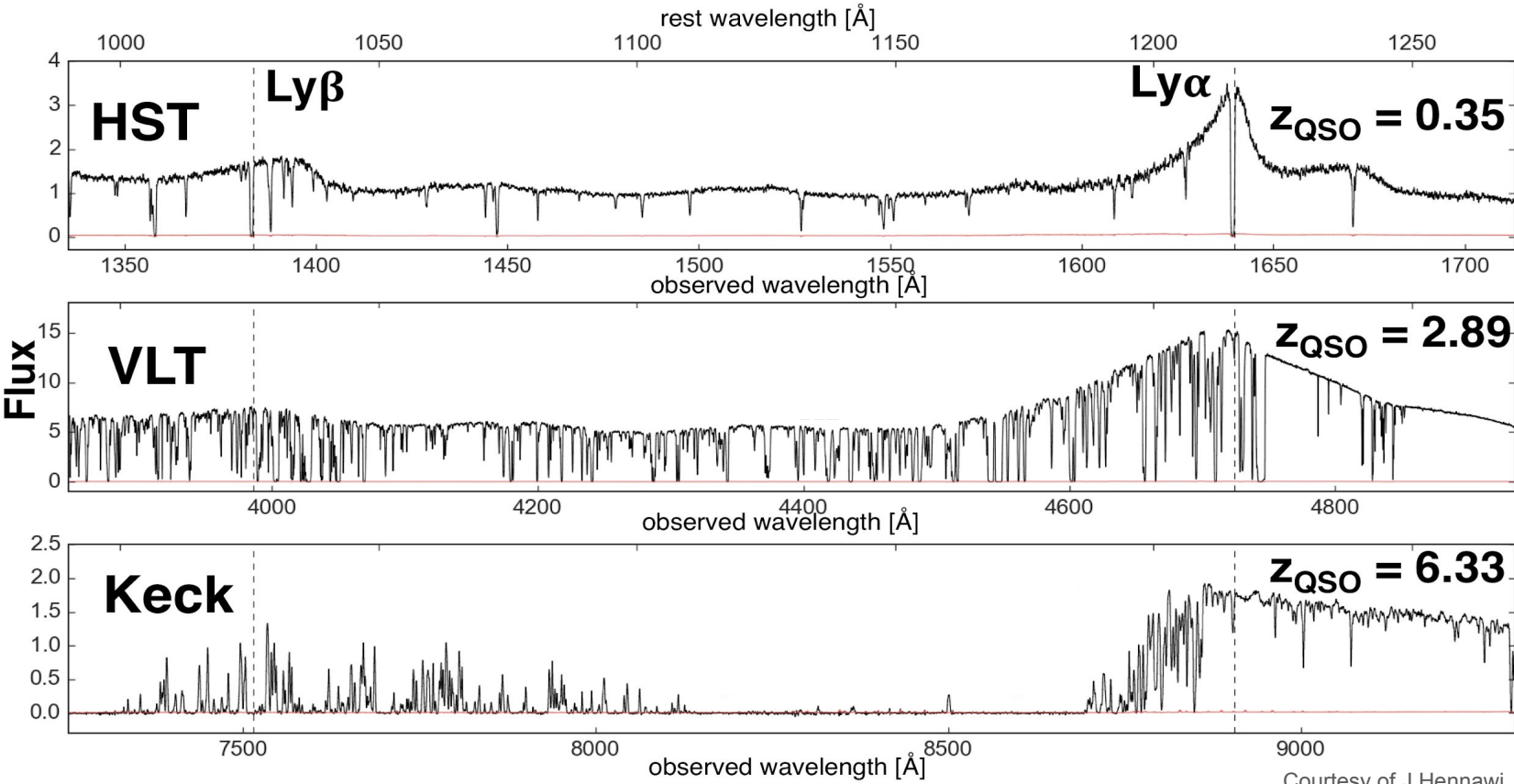


Christina Eilers

# Probing the IGM with the Lyman- $\alpha$ forest



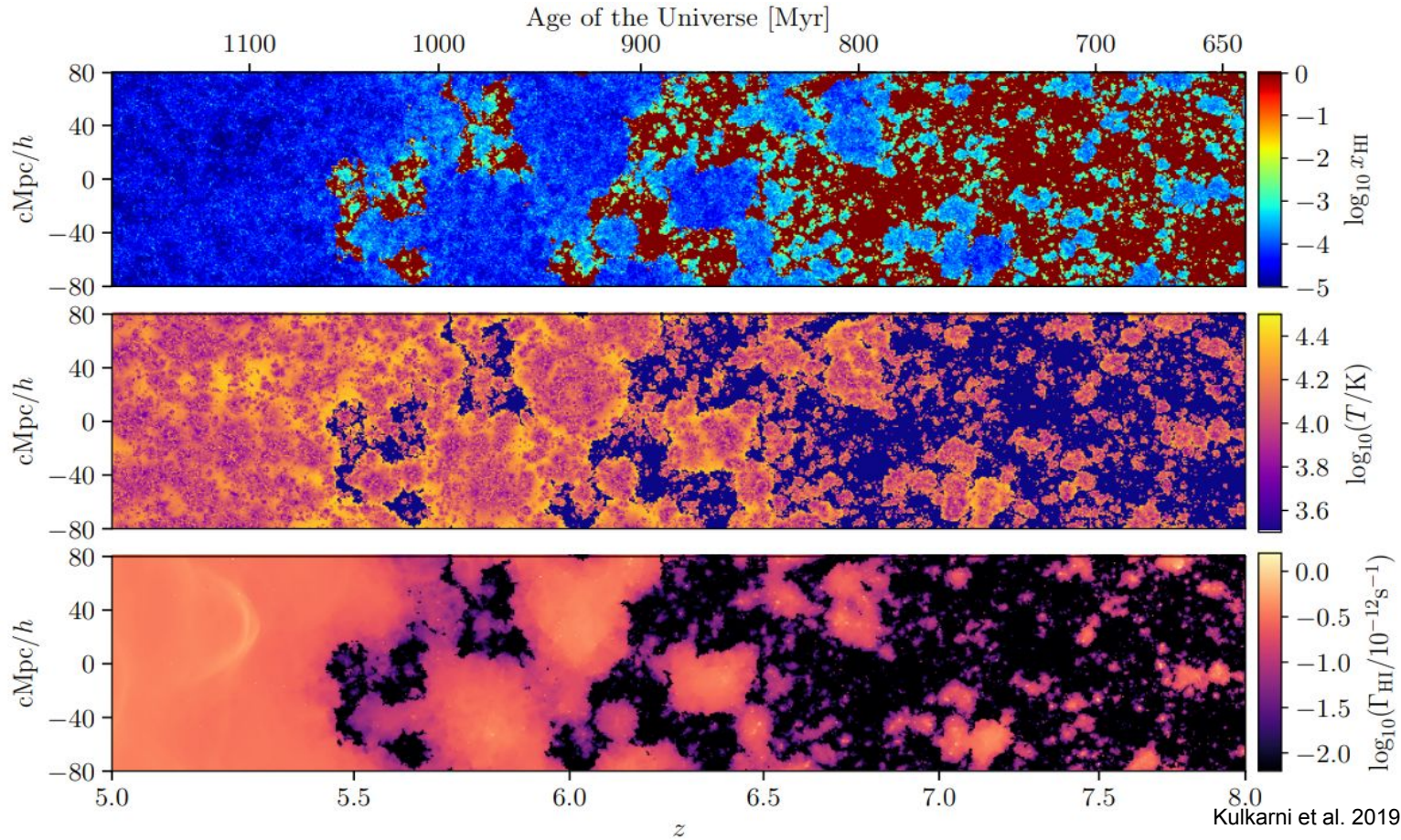
# Probing the IGM with the Lyman- $\alpha$ forest



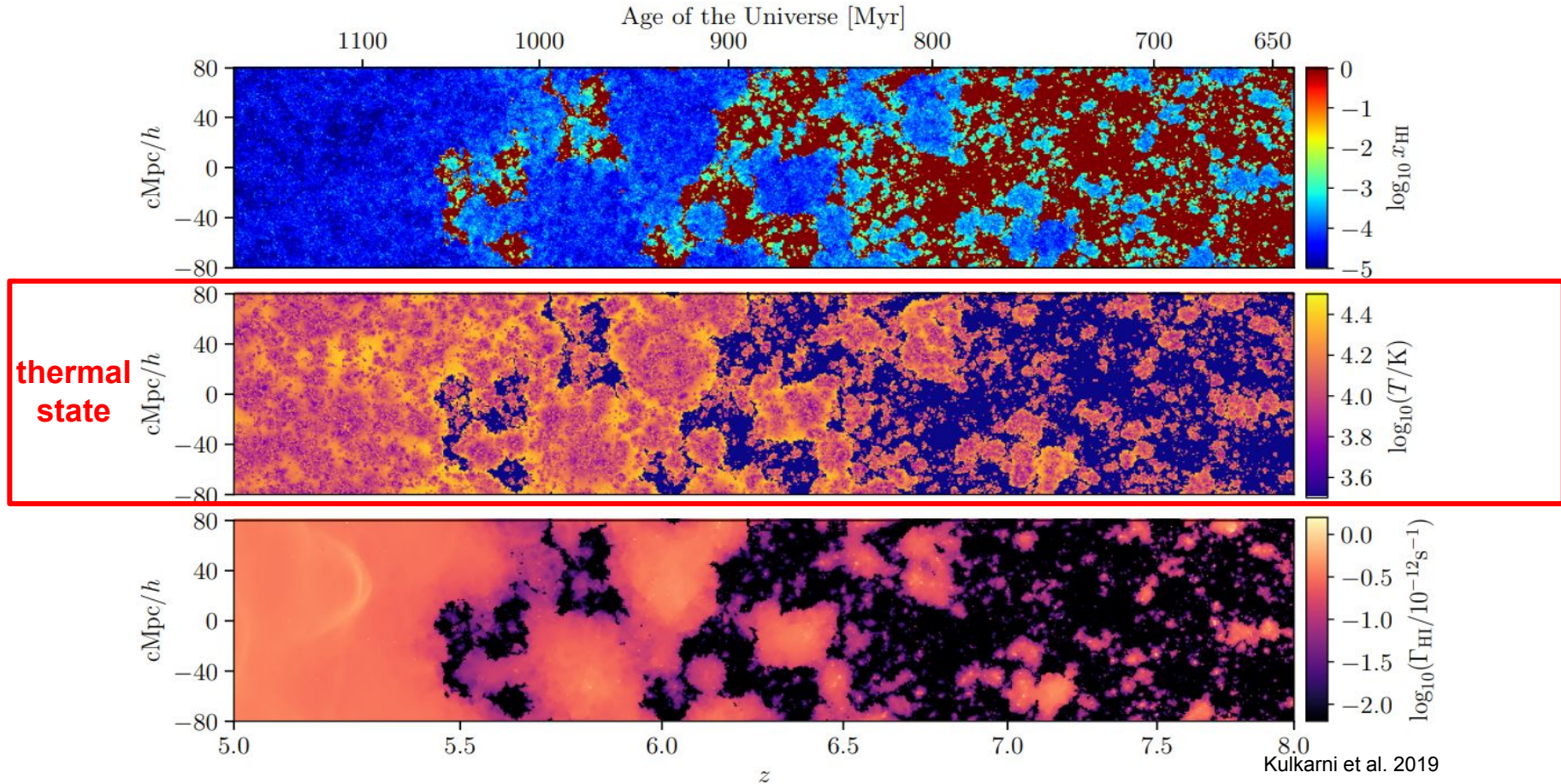
Saturates for  $\langle x_{\text{HI}}(z) \rangle \gtrsim 10^{-4}$



# Some signatures of reionization in the IGM

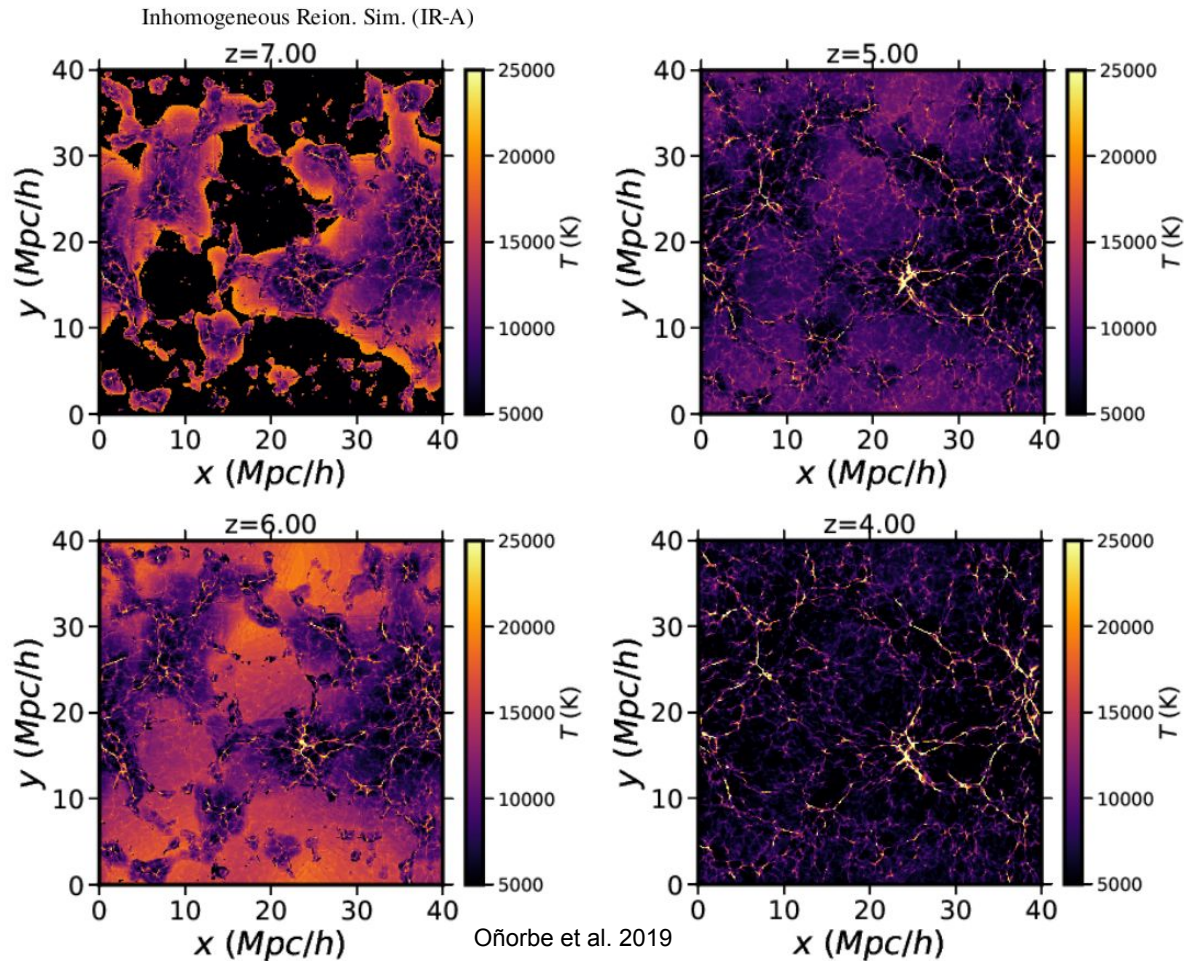


# Some signatures of reionization in the IGM

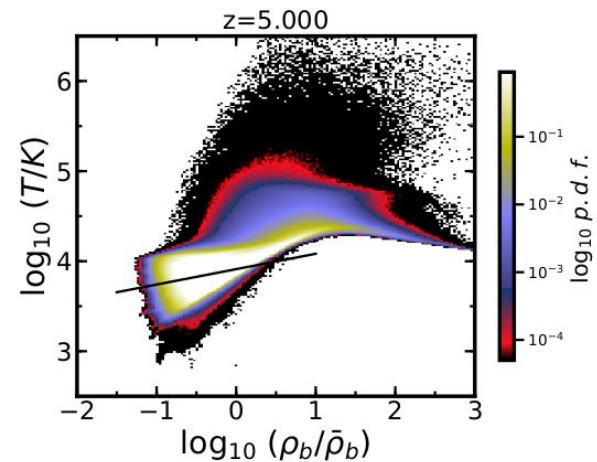




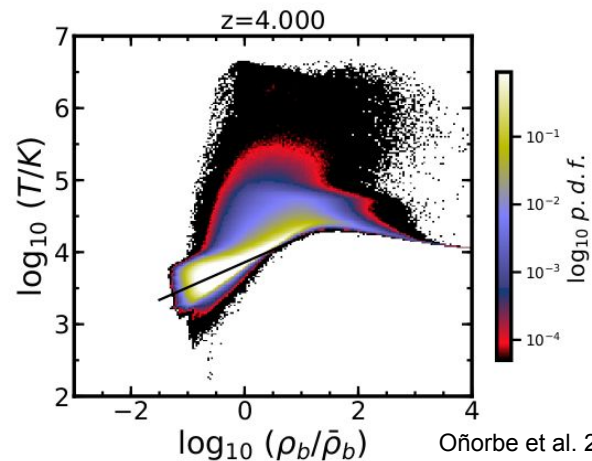
# Thermal state evolution in simulations



# Parameters for the thermal state

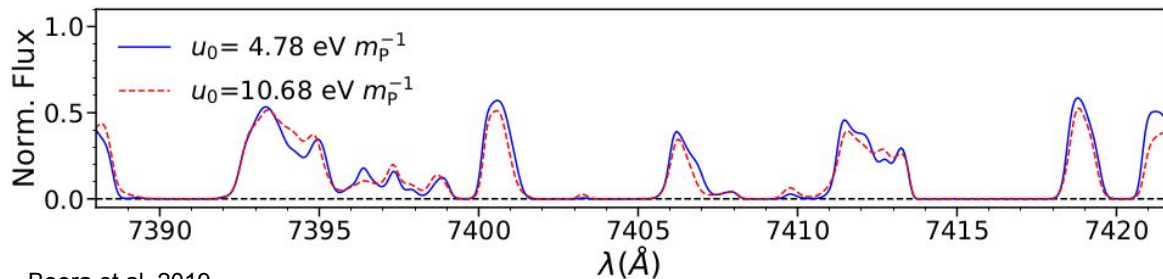
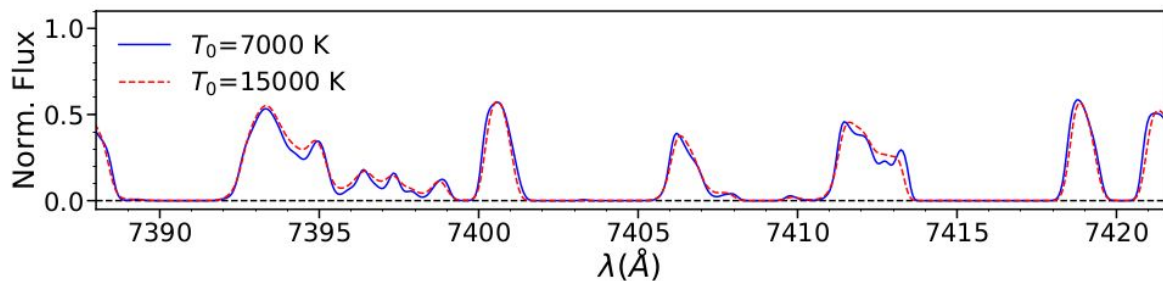


$$T = T_0(\rho/\bar{\rho})^{\gamma-1}$$



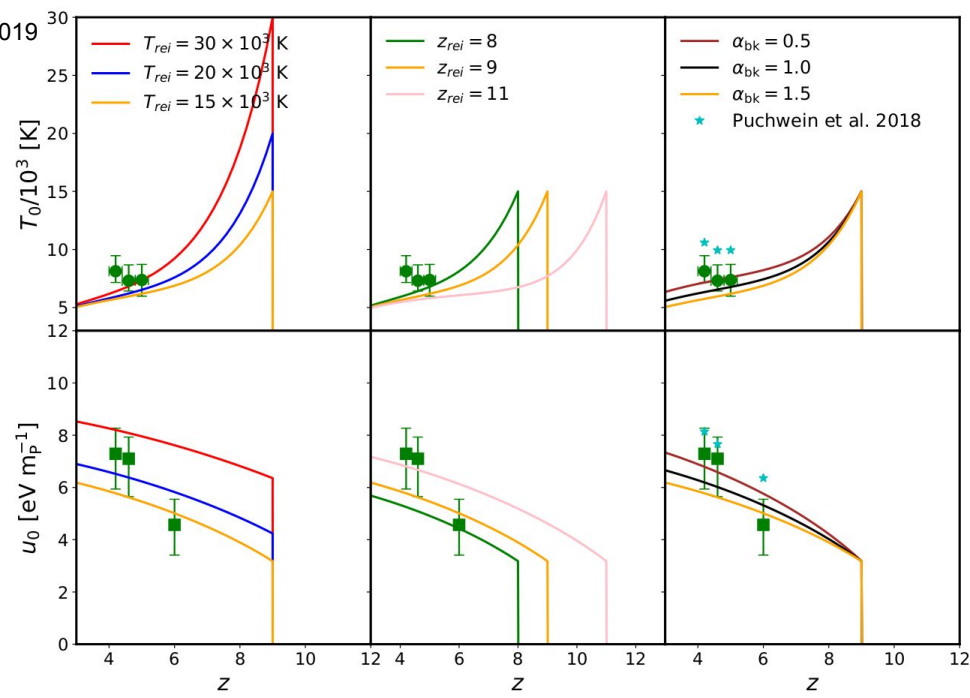
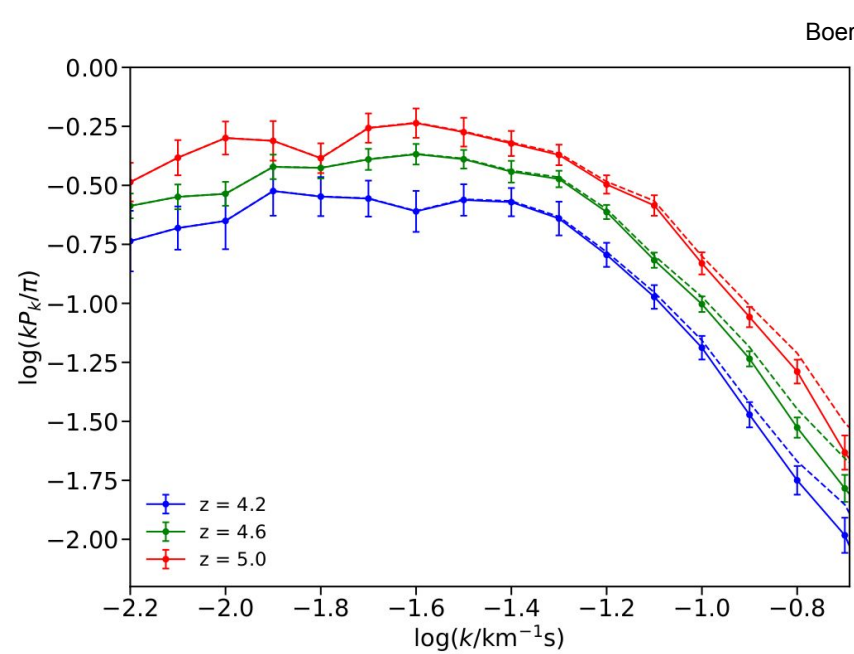
Oñorbe et al. 2019

Also pressure smoothing (here  $u_0$ )



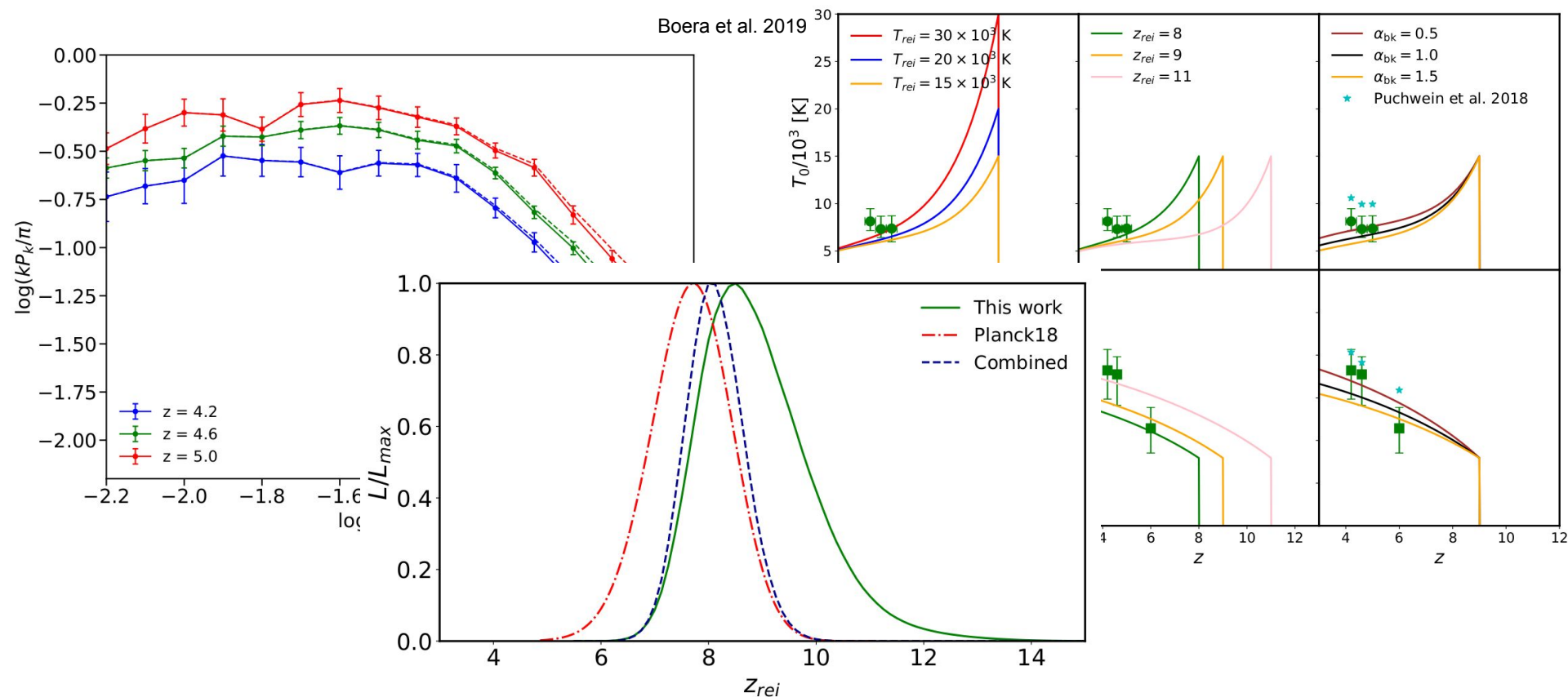
Boera et al. 2019

# Existing measurements of the thermal state

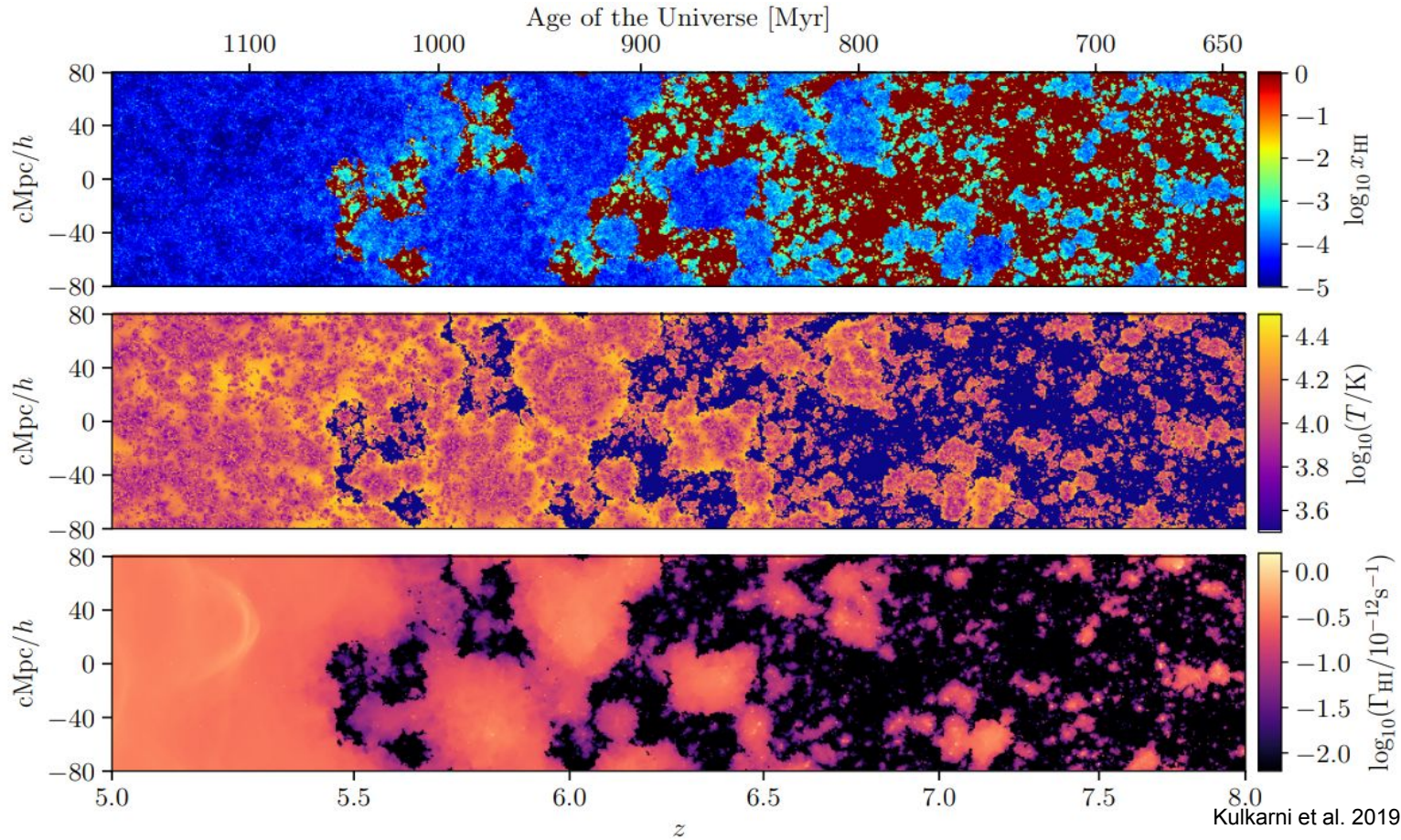




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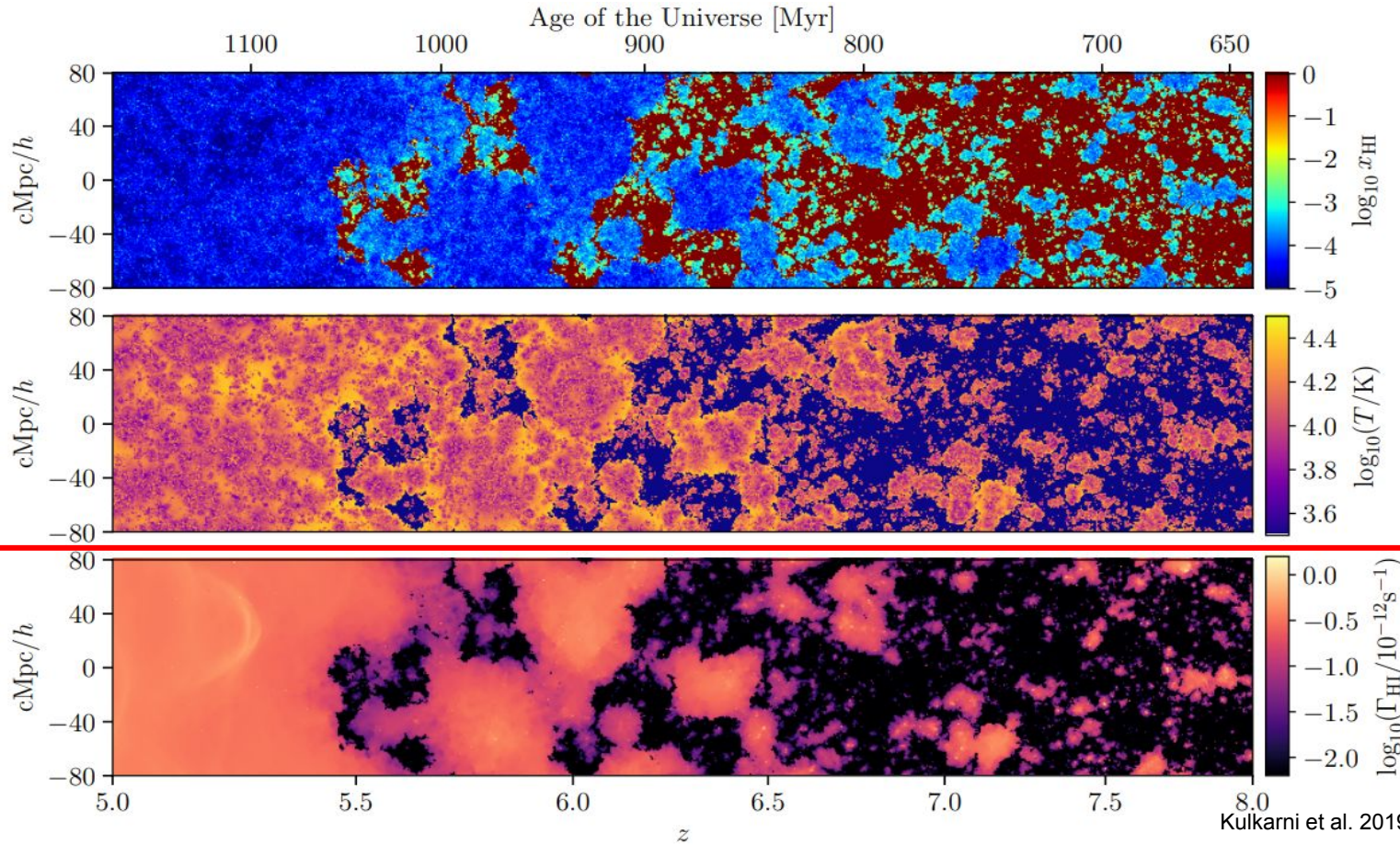


# Some signatures of reionization in the IGM



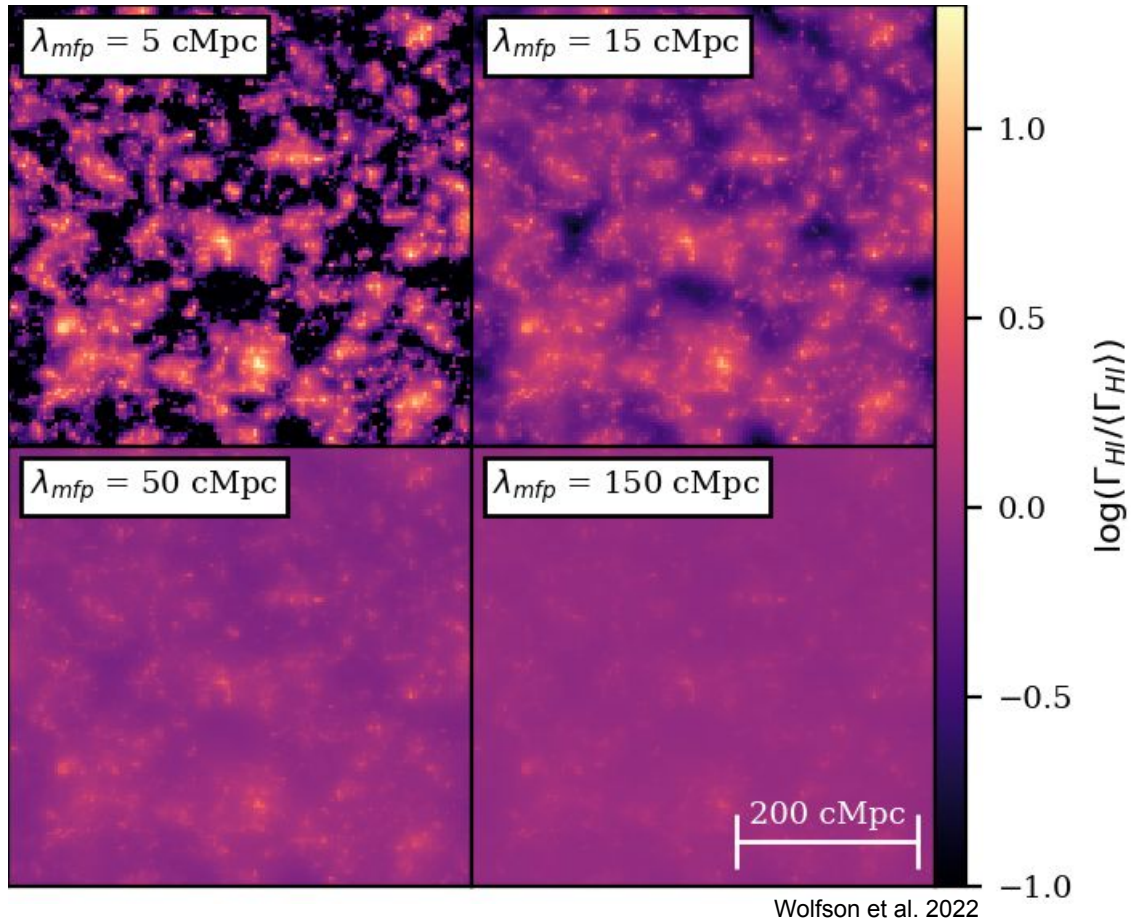


# Some signatures of reionization in the IGM



Fluctuations in the UVB can be described by  $\lambda_{\text{mfp}}$

$\lambda_{\text{mfp}}$  - the average  
distance ionizing  
photons travel before  
interacting with neutral  
hydrogen

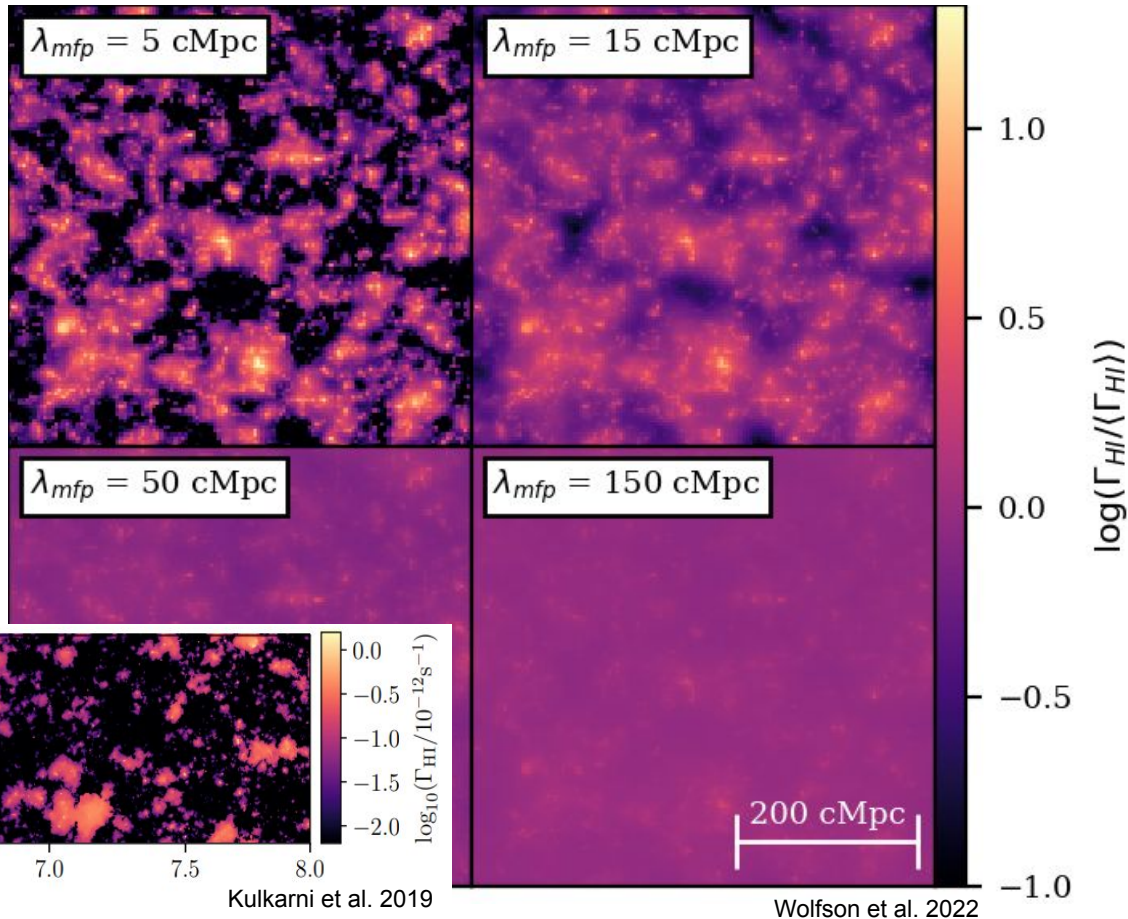
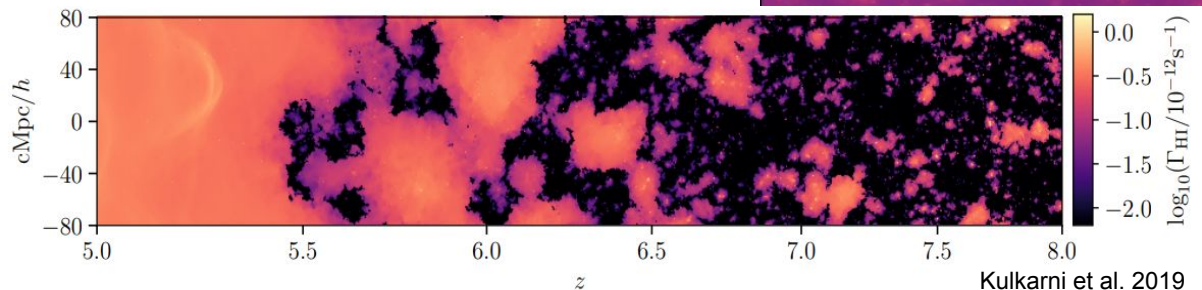




# Fluctuations in the UVB can be described by $\lambda_{\text{mfp}}$

$\lambda_{\text{mfp}}$  - the average distance ionizing photons travel before interacting with neutral hydrogen

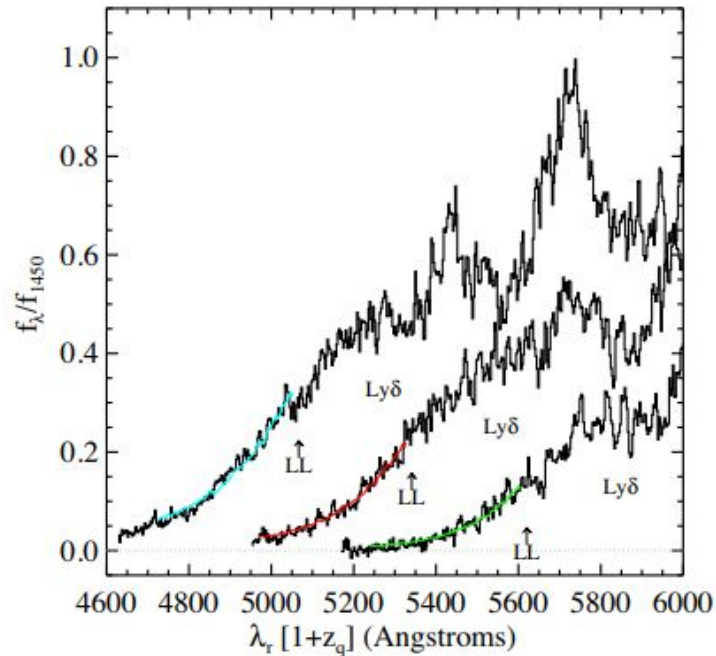
Rapid increase signals the end of reionization



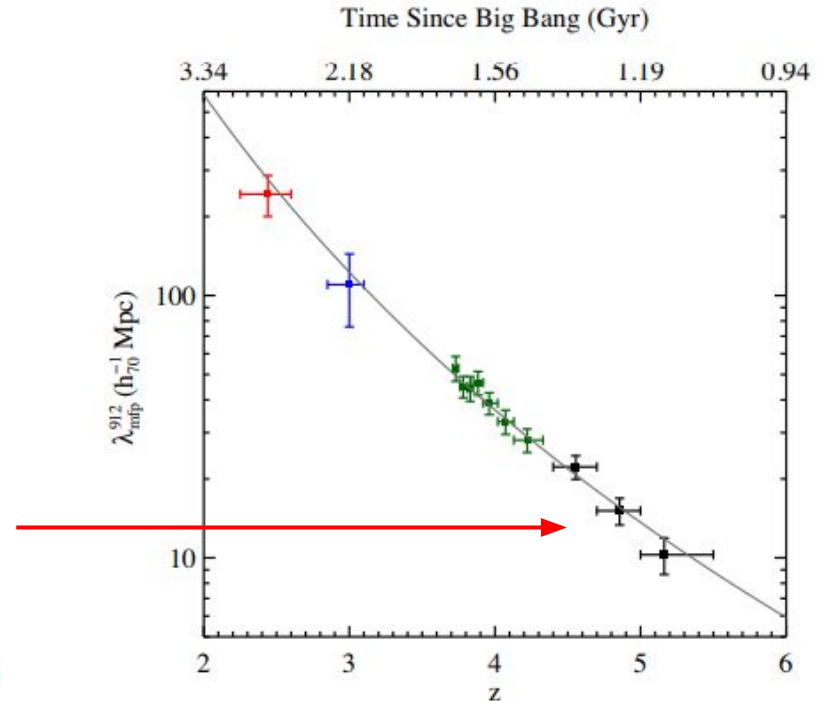


# Existing measurements of $\lambda_{\text{mfp}}$

Most constraining method has been from flux beyond the Lyman limit in stacked quasar spectra



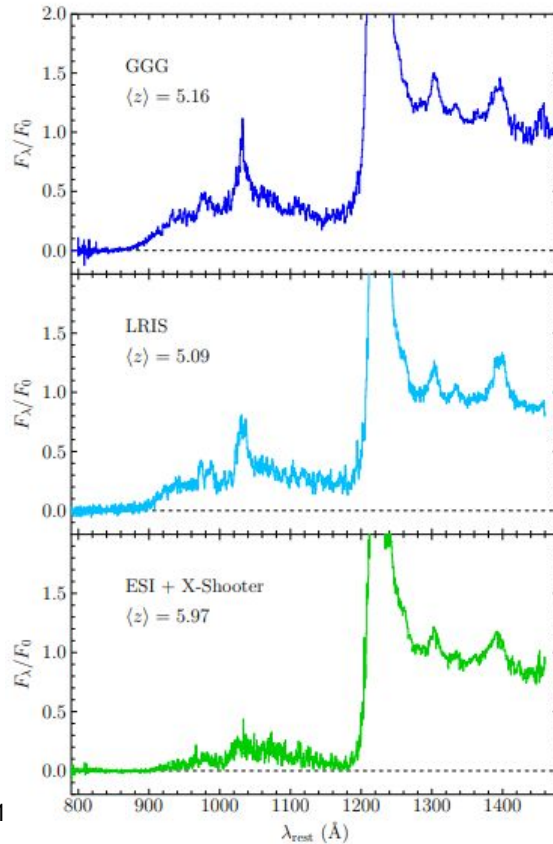
Worseck et al. 2014



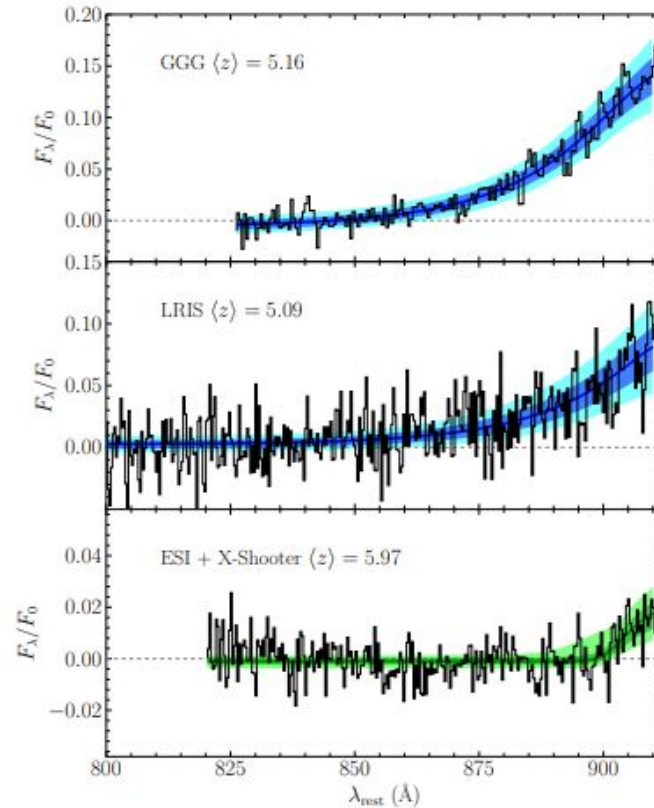
Worseck et al. 2014

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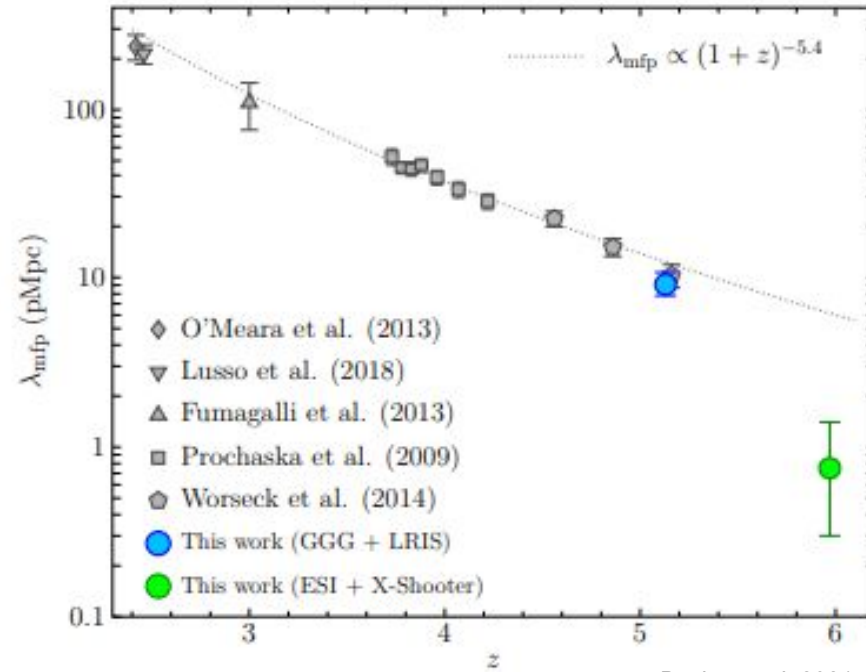
Becker et al. 2021



Becker et al. 2021

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Becker et al. 2021

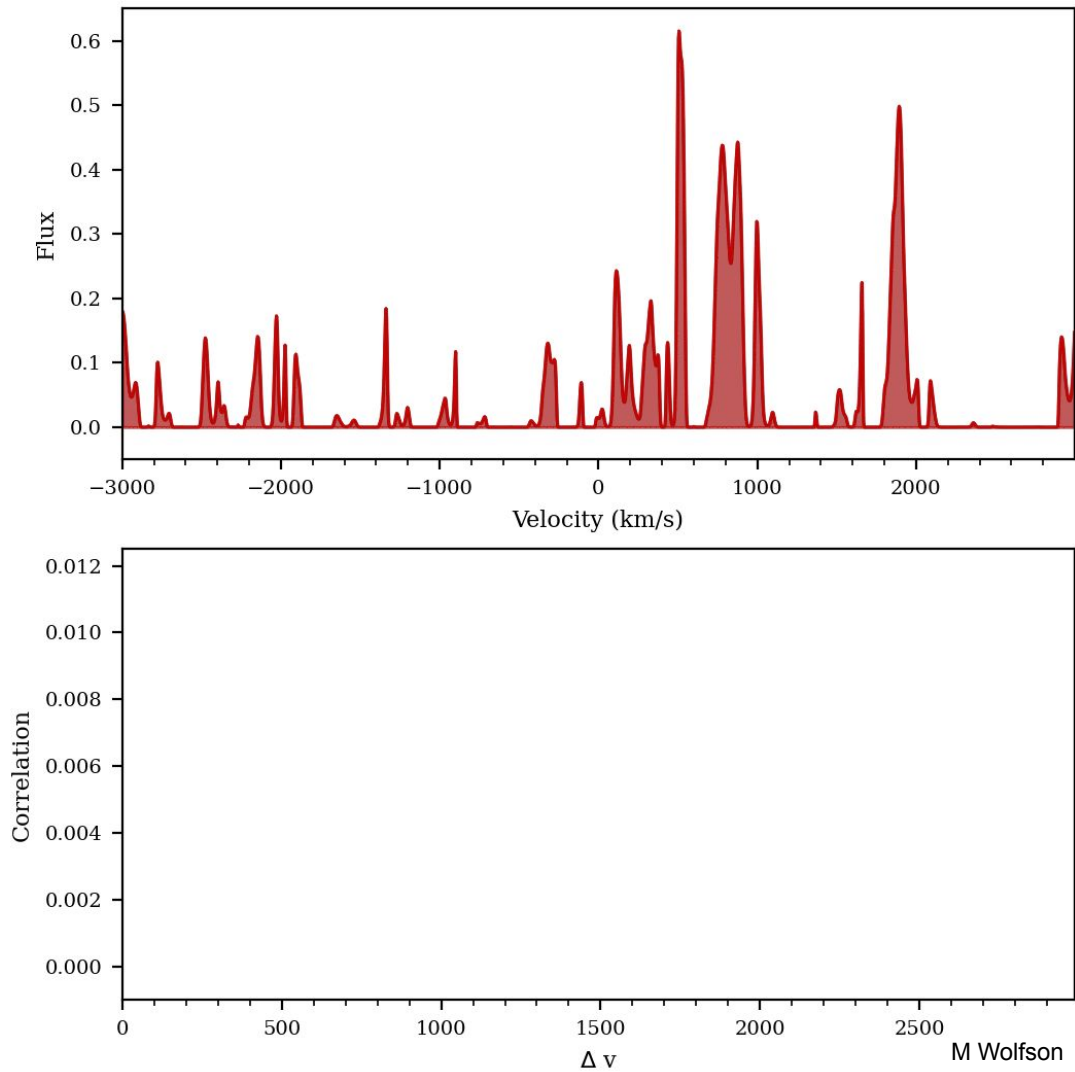
# Auto-correlation function

$$\xi_F(\Delta v) = \langle F(v) F(v + \Delta v) \rangle$$

The fourier transform of the power spectrum

Uncorrelated gaussian noise averages out

Easy to mask out DLAs etc



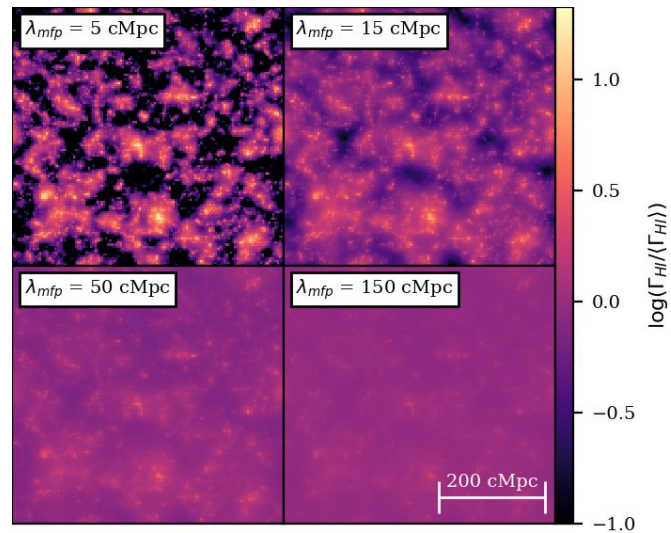
Is the auto-correlation function of the  
Lyman- $\alpha$  forest sensitive to  $\lambda_{\text{mfp}}$ ?



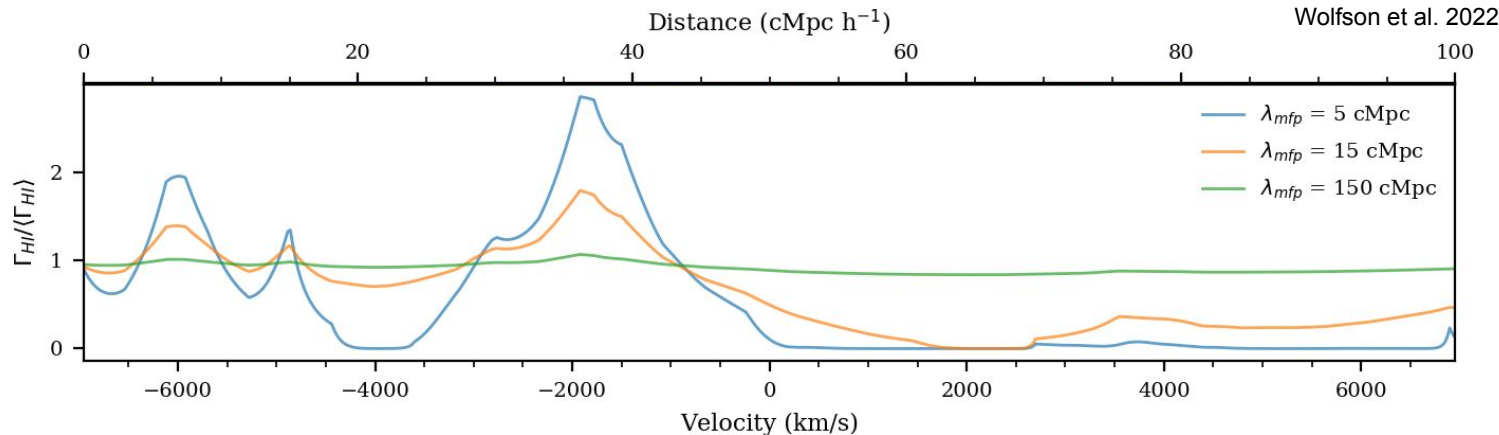
# Simulation box properties

UVB boxes with  $L_{\text{box}} = 512 \text{ cMpc}$

- Method of Davies & Furlanetto 2016
- $128^3$  pixels



Wolfson et al. 2022



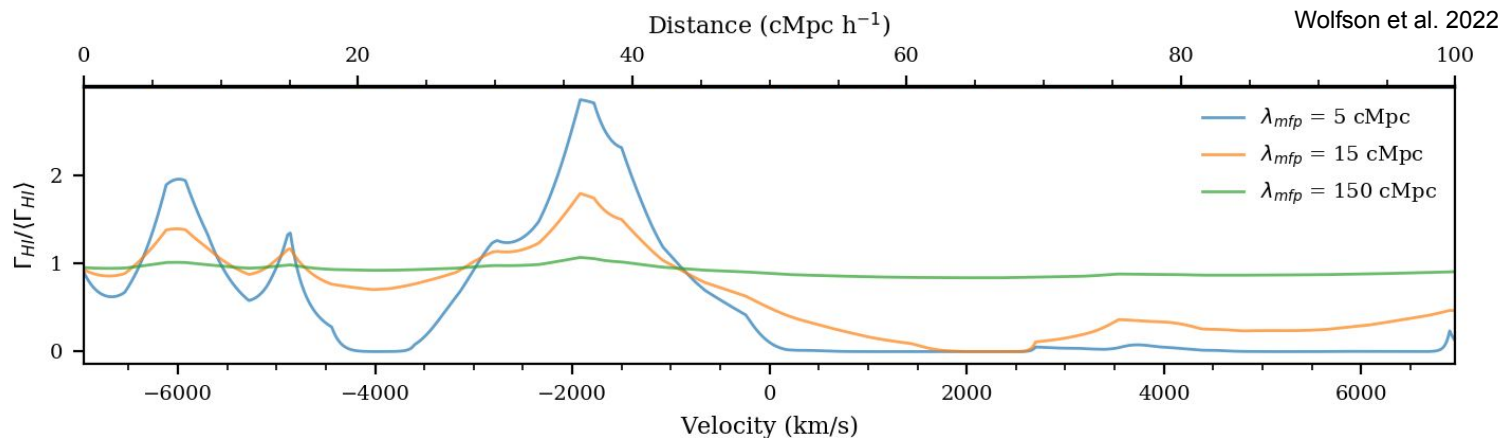
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Nyx box with  $L_{\text{box}} = 100 \text{ cMpc } h^{-1}$

- Hydrodynamical code designed for the Ly $\alpha$  forest
- $4096^3$  dark matter particles,  $4096^3$  baryon grid cells



# Simulation box properties

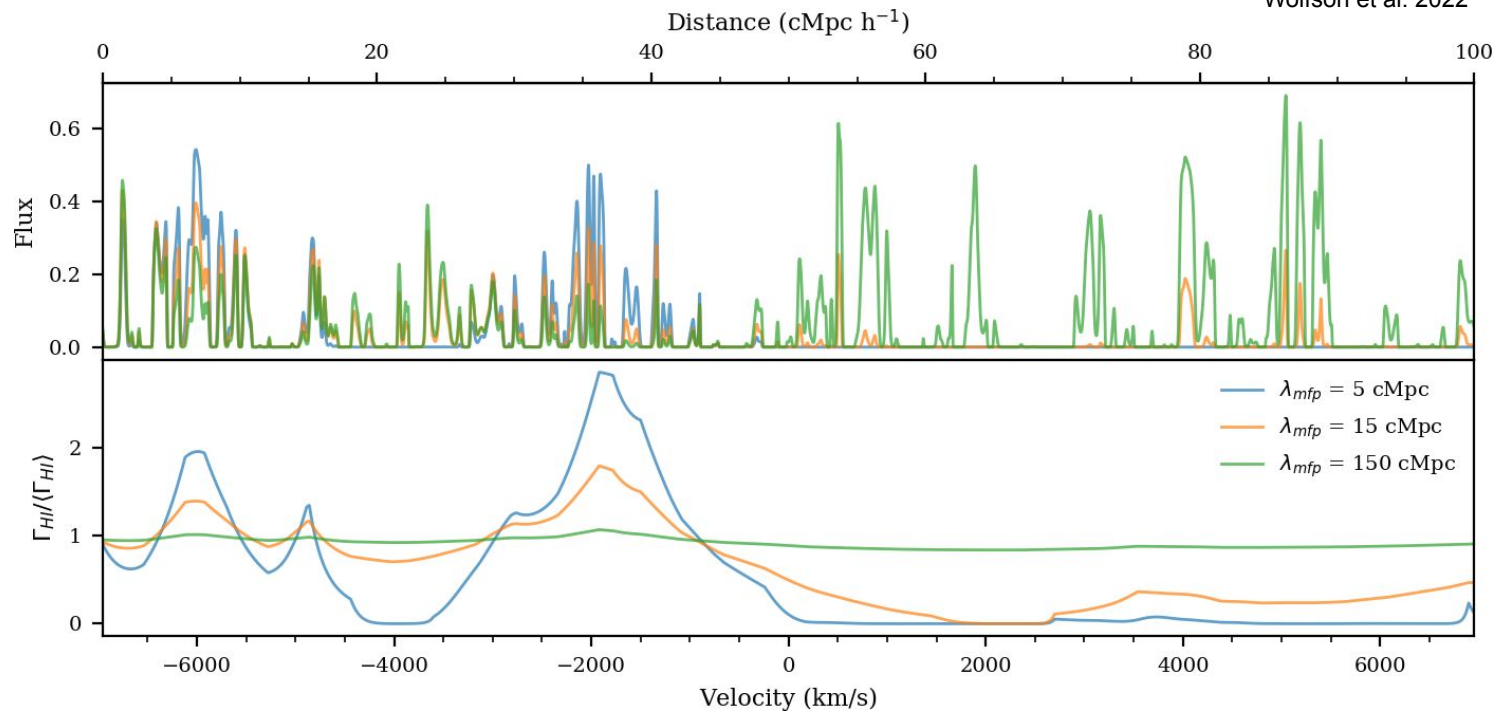
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Wolfson et al. 2022



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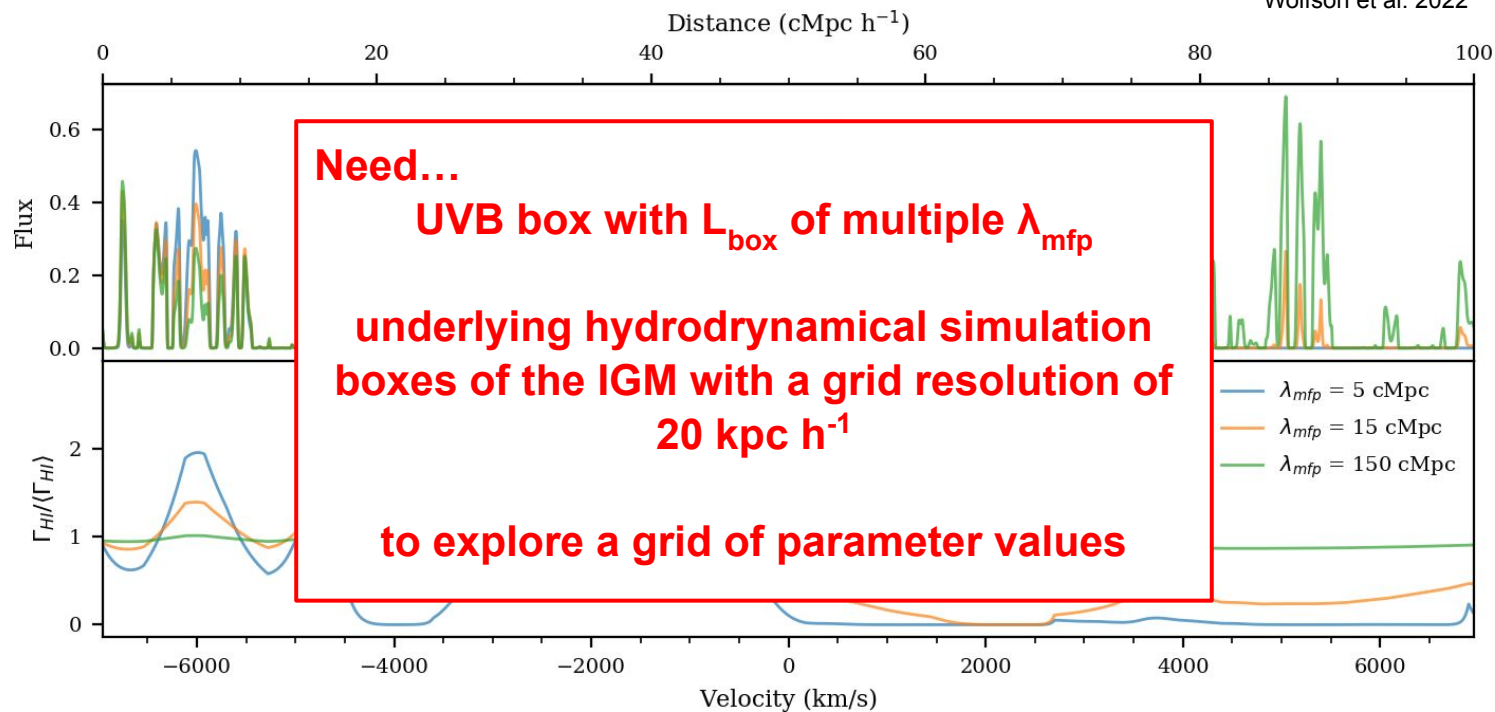
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Wolfson et al. 2022

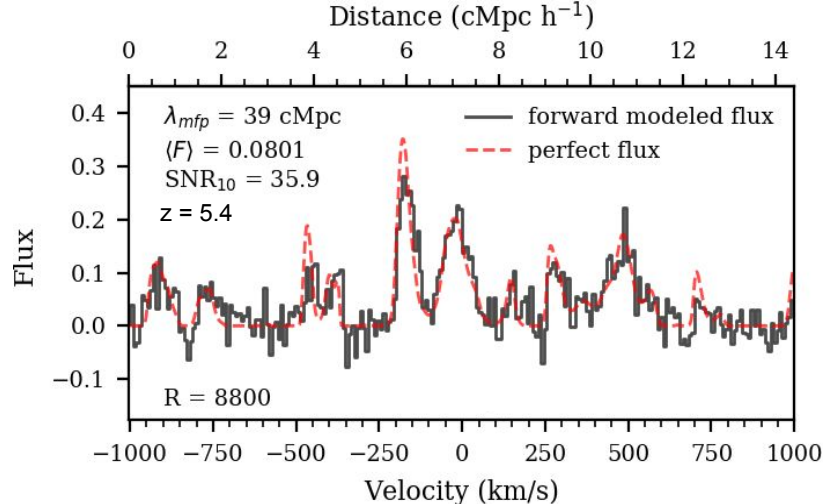


# Forward modeling realistic data

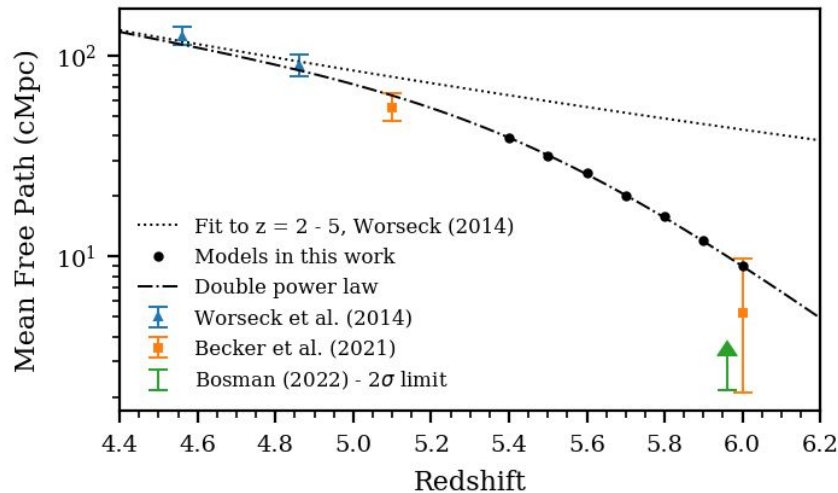
## Forward modeling properties:

- Resolution of 8800
  - The resolution of X-Shooter
  - Convolve with a gaussian filter
- Gaussian noise with  $\text{SNR}_{10} = 35.9$ 
  - Median of data in Bosman et al. 2021
- 12-60 quasars in each redshift bin

Double power law - fit by eye - of the evolution of  $\lambda_{\text{mfp}}$

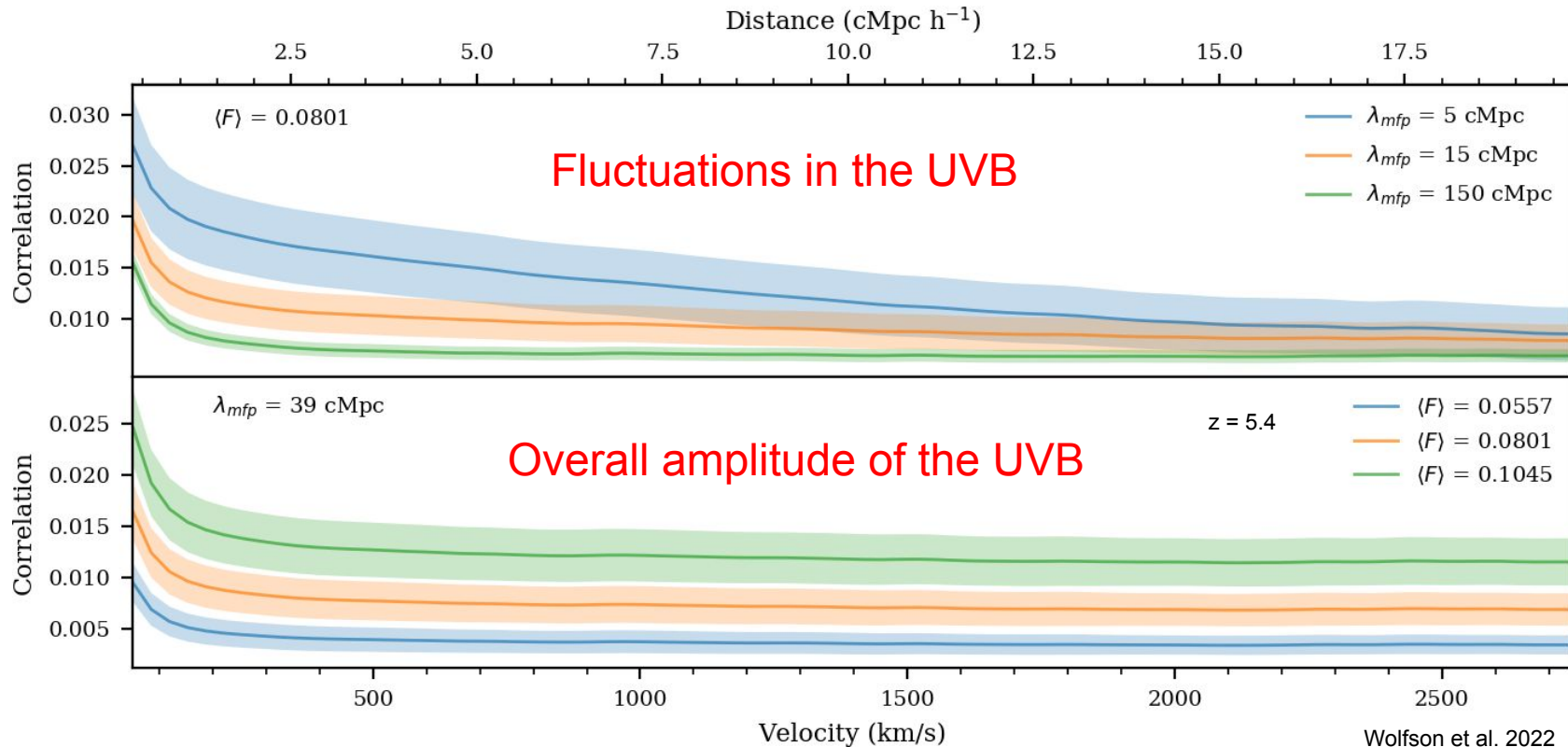


Wolfson et al. 2022





# Effect of $\lambda_{\text{mfp}}$ on the auto-correlation function

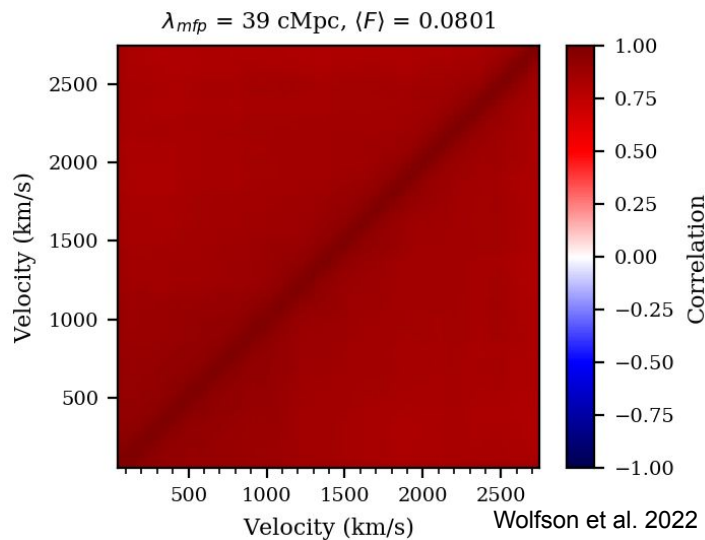


Differences between the models for different  $\lambda_{\text{mfp}}$  values are non-linear!

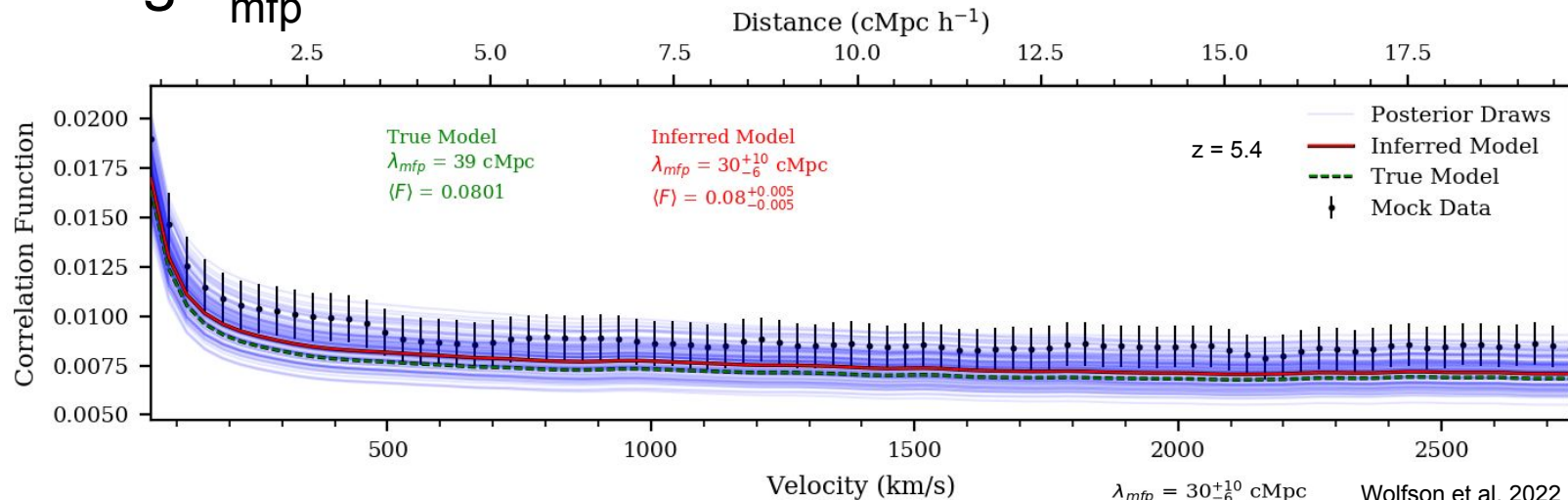
# Measuring $\lambda_{\text{mfp}}$ from mock data

Gaussian likelihood:

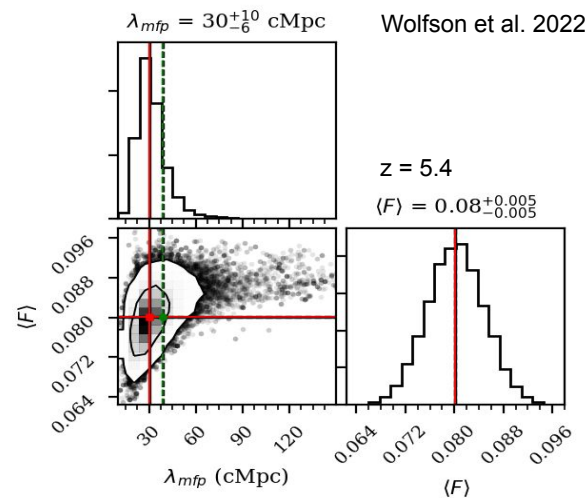
$$\mathcal{L} = \frac{1}{\sqrt{\det(\Sigma)}(2\pi)^n} \exp \left( -\frac{1}{2} (\boldsymbol{\xi} - \boldsymbol{\xi}_{\text{model}}(\lambda_{\text{mfp}}, \langle F \rangle))^T \Sigma^{-1} (\boldsymbol{\xi} - \boldsymbol{\xi}_{\text{model}}(\lambda_{\text{mfp}}, \langle F \rangle)) \right)$$



# Measuring $\lambda_{\text{mfp}}$ from mock data



MCMC



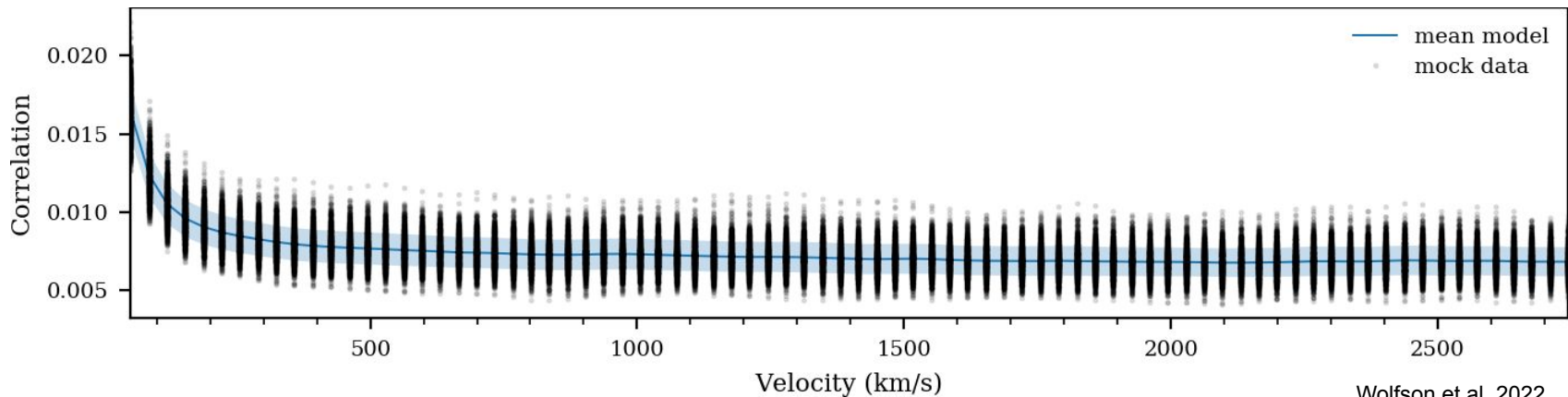
Can we check that our assumptions are correct?

Do we know our posteriors accurately reflect the true probabilities of our measurements?

# Is the assumption of gaussian distributed data correct?

$$\mathcal{L} = \frac{1}{\sqrt{\det(\Sigma)(2\pi)^n}} \exp \left( -\frac{1}{2} (\boldsymbol{\xi} - \boldsymbol{\xi}_{\text{model}}(\lambda_{\text{mfp}}, \langle F \rangle))^T \Sigma^{-1} (\boldsymbol{\xi} - \boldsymbol{\xi}_{\text{model}}(\lambda_{\text{mfp}}, \langle F \rangle)) \right)$$

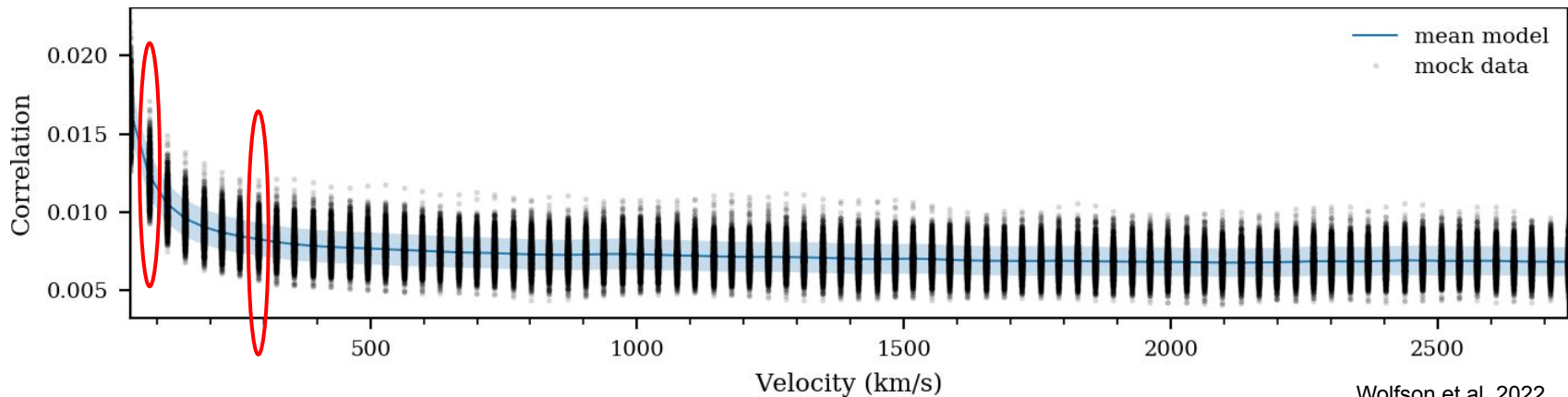
- Flux PDF is not gaussian but assume averaging to auto-correlation gaussianizes data
- Commonly done with weak lensing, other Ly $\alpha$  forest statistics



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$$\mathcal{L} = \frac{1}{\sqrt{\det(\Sigma)}(2\pi)^n} \exp \left( -\frac{1}{2} (\boldsymbol{\xi} - \boldsymbol{\xi}_{\text{model}}(\lambda_{\text{mfp}}, \langle F \rangle))^T \Sigma^{-1} (\boldsymbol{\xi} - \boldsymbol{\xi}_{\text{model}}(\lambda_{\text{mfp}}, \langle F \rangle)) \right)$$

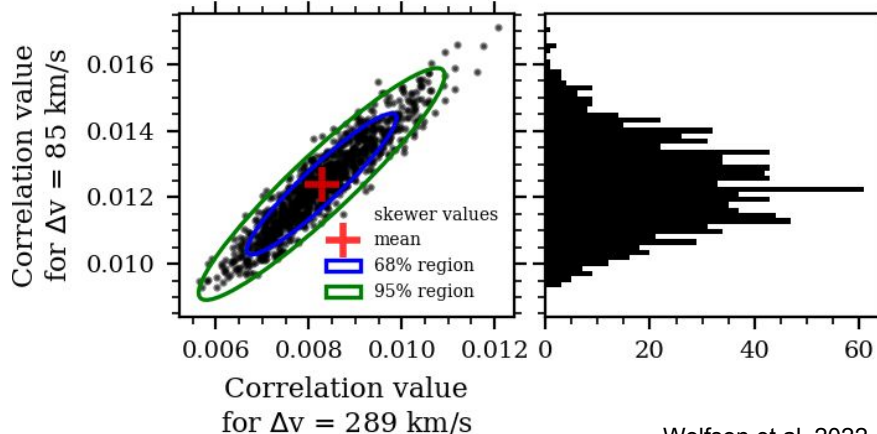
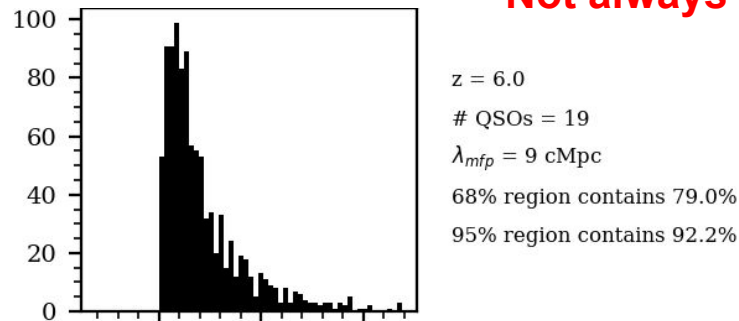
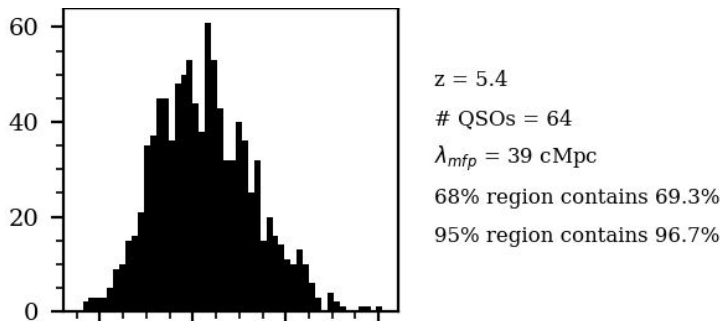
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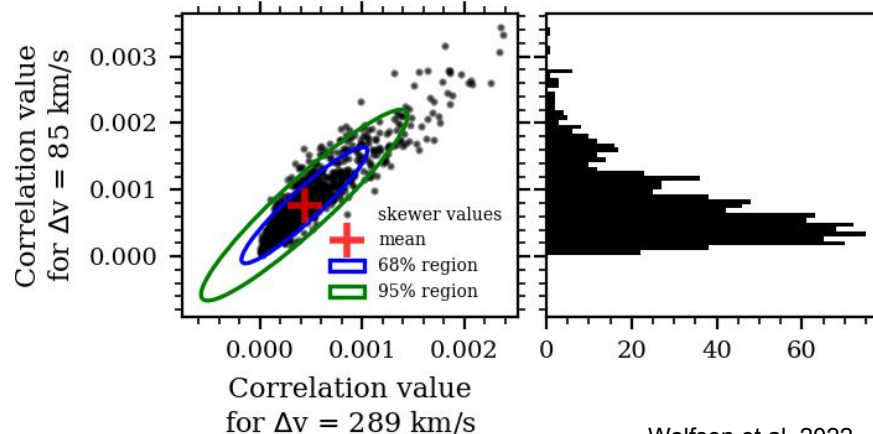


# Is the assumption of gaussian distributed data correct?

**Not always**



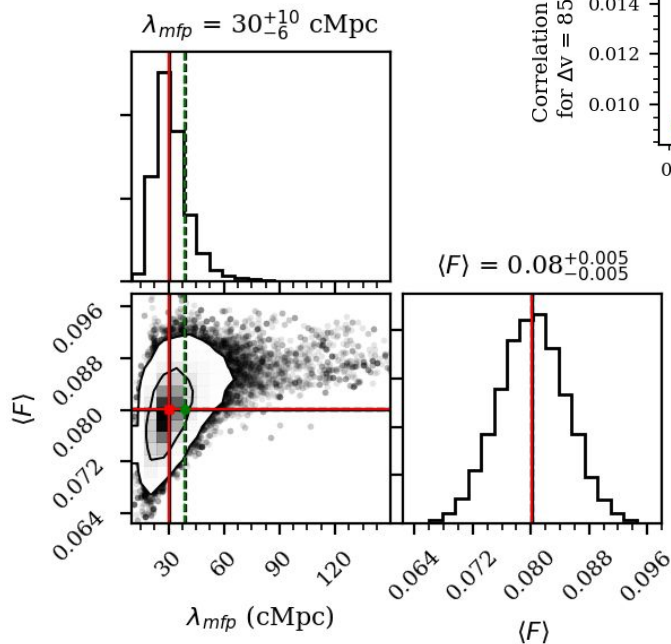
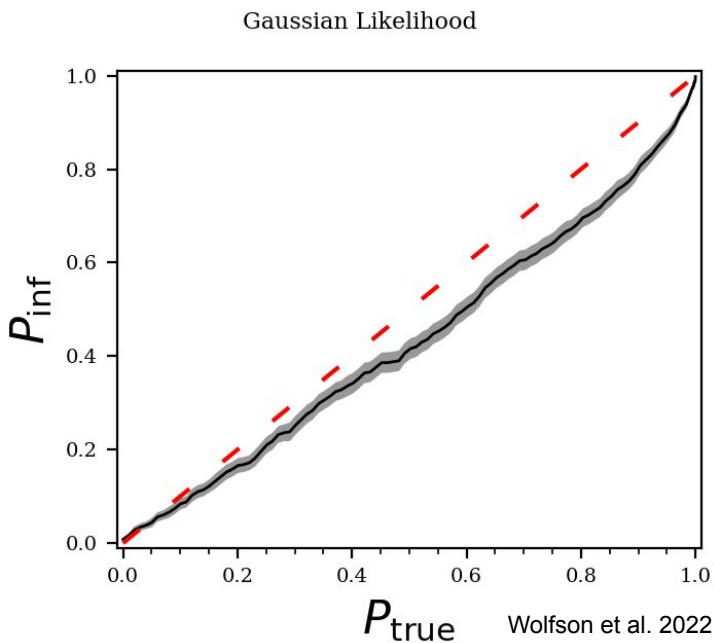
Wolfson et al. 2022



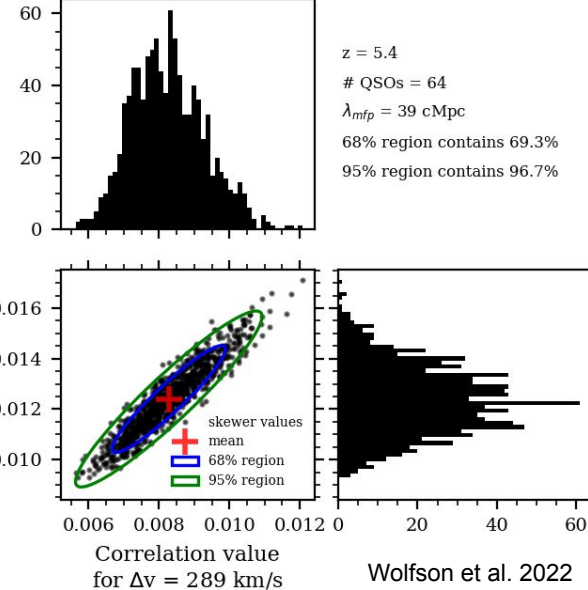
Wolfson et al. 2022

# Can we quantify how correct our assumptions are?

Look at the posteriors for 500 mock data sets

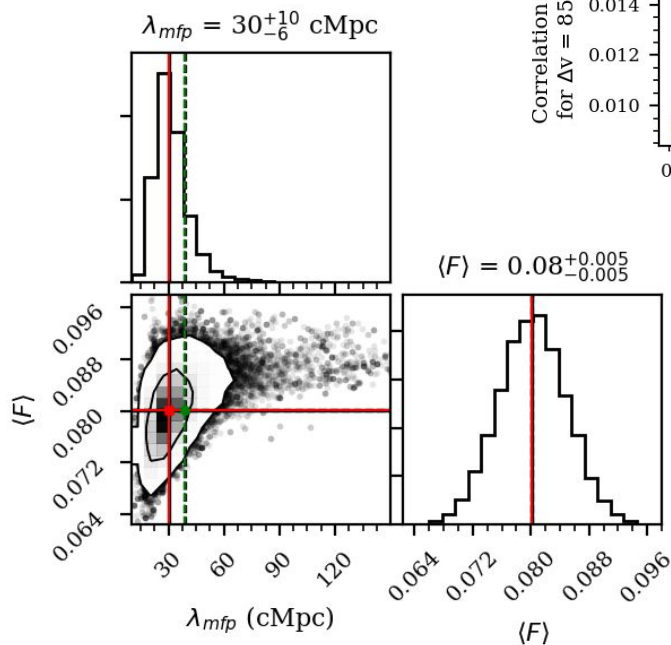
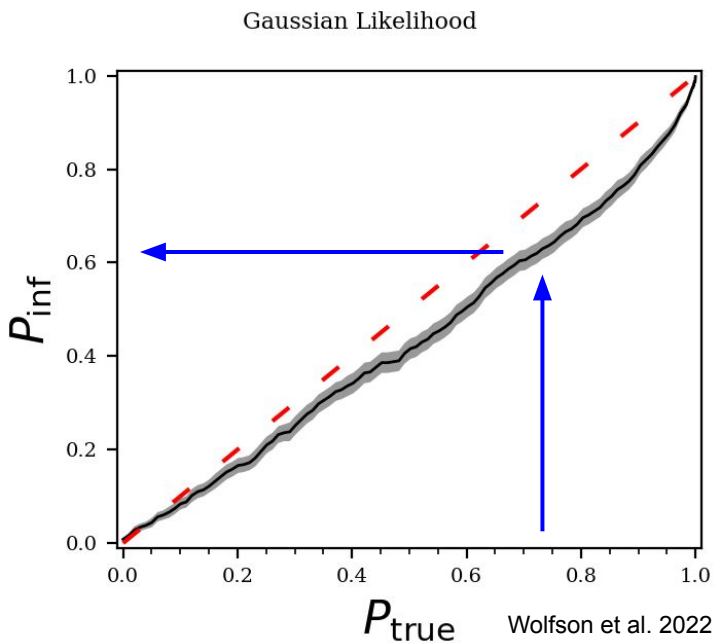


Wolfson et al. 2022

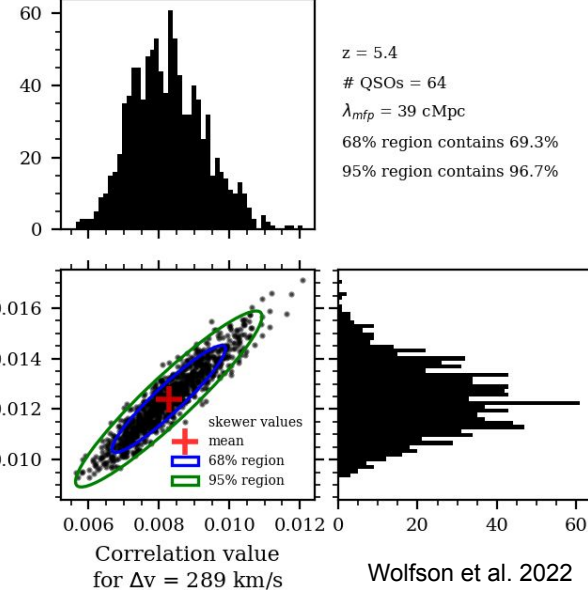


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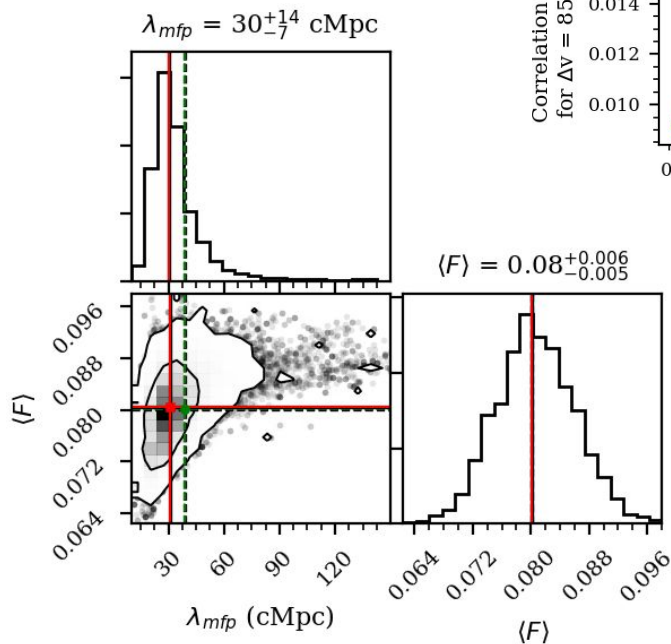
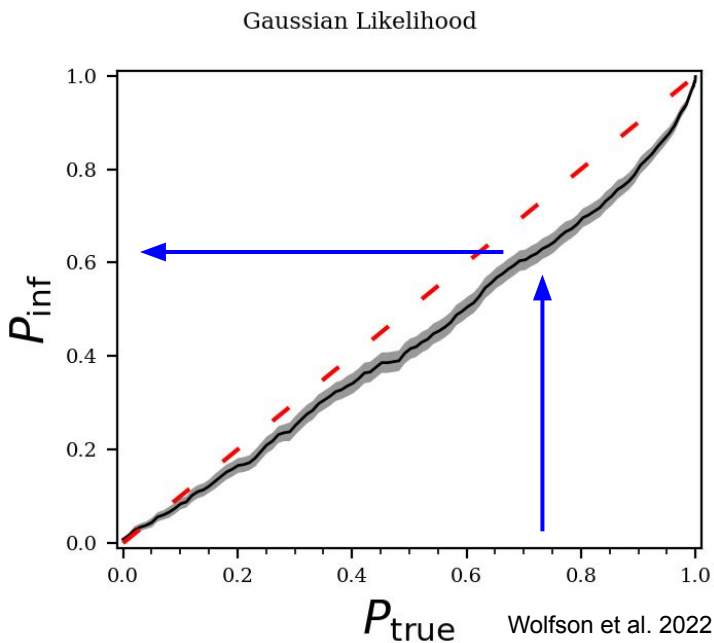
Wolfson et al. 2022



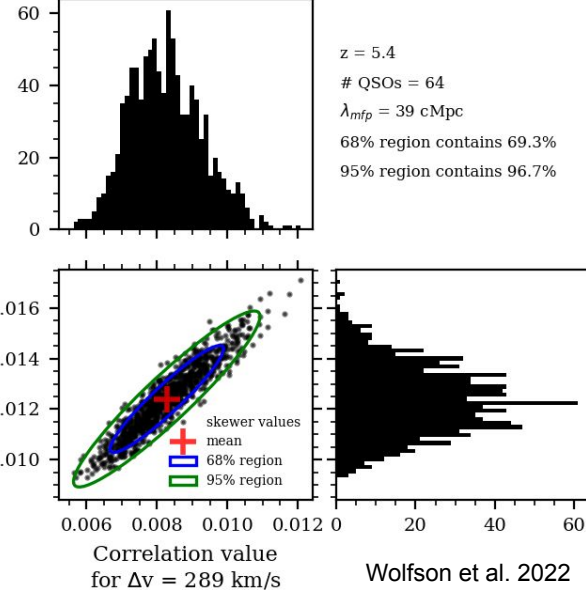
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Look at the posteriors for 500 mock data sets

Calculate one set of weights for all posteriors to pass



Wolfson et al. 2022

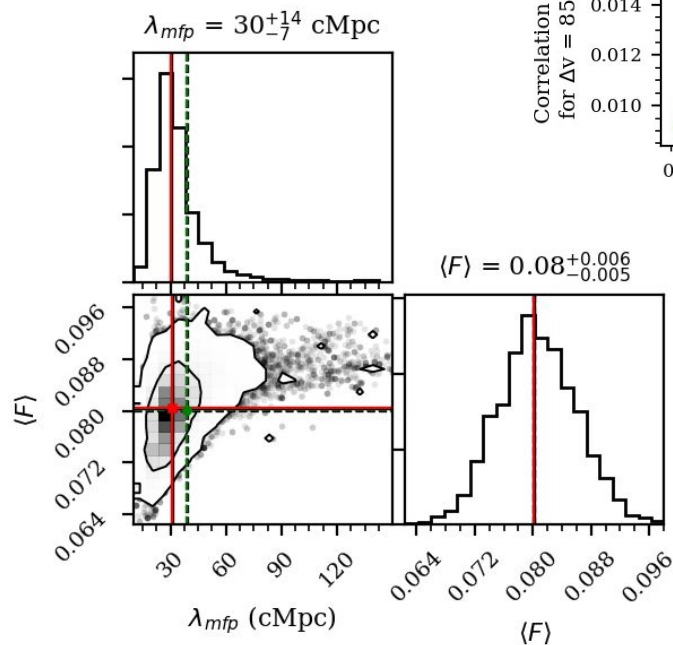
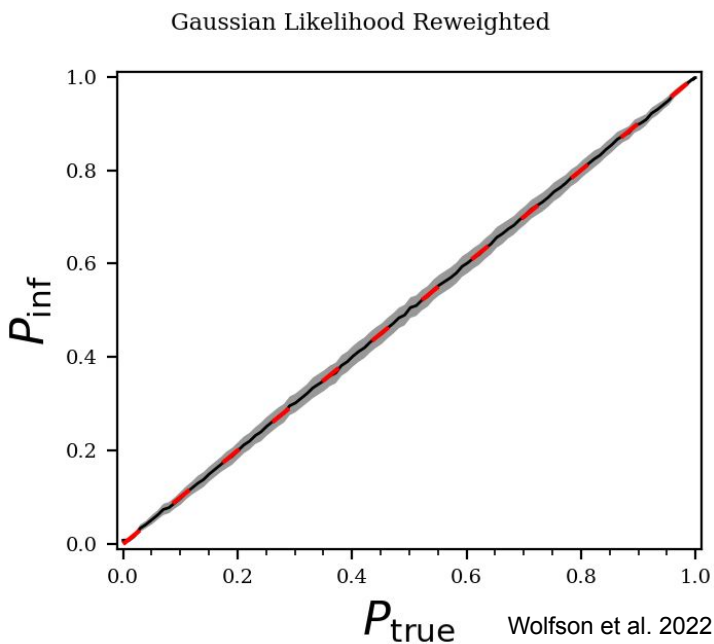




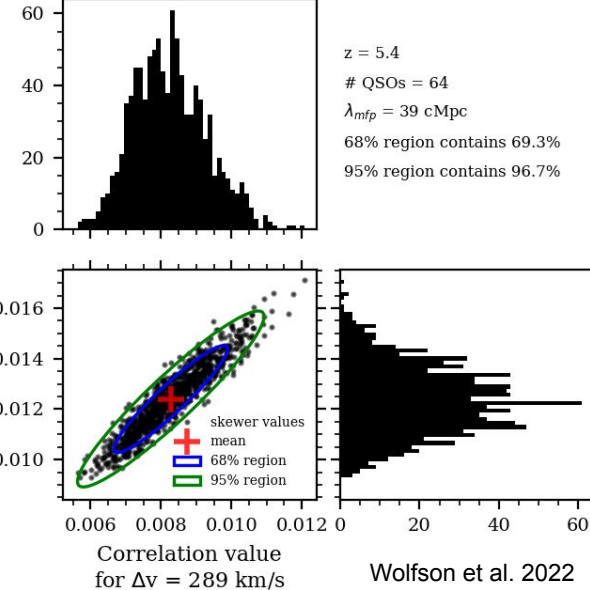
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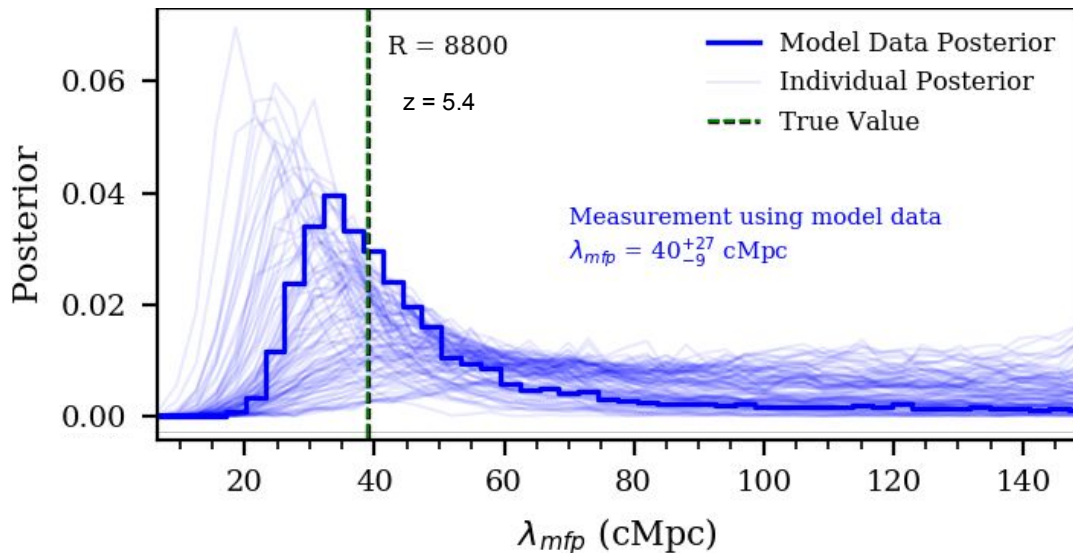


Wolfson et al. 2022



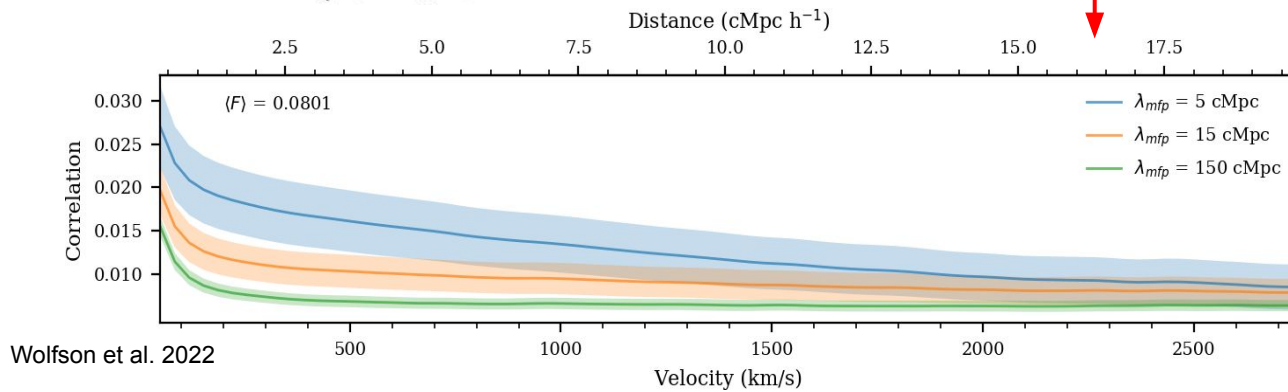
# Measuring $\lambda_{\text{mfp}}$ from mock data

Wolfson et al. 2022



Luck of the draw for the chosen mock data set dictates the precision of the posterior

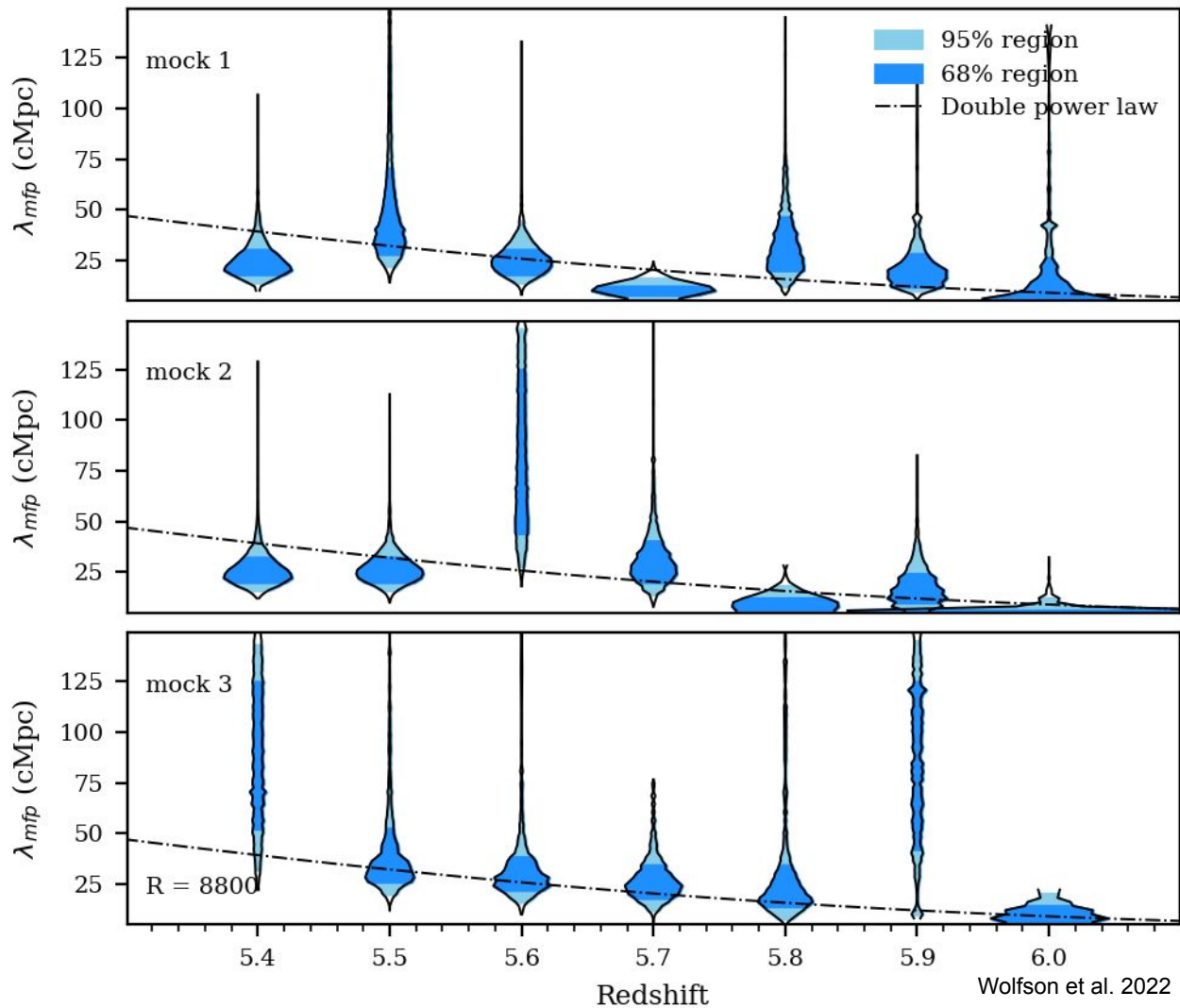
Remove this by using the **model value** as the data



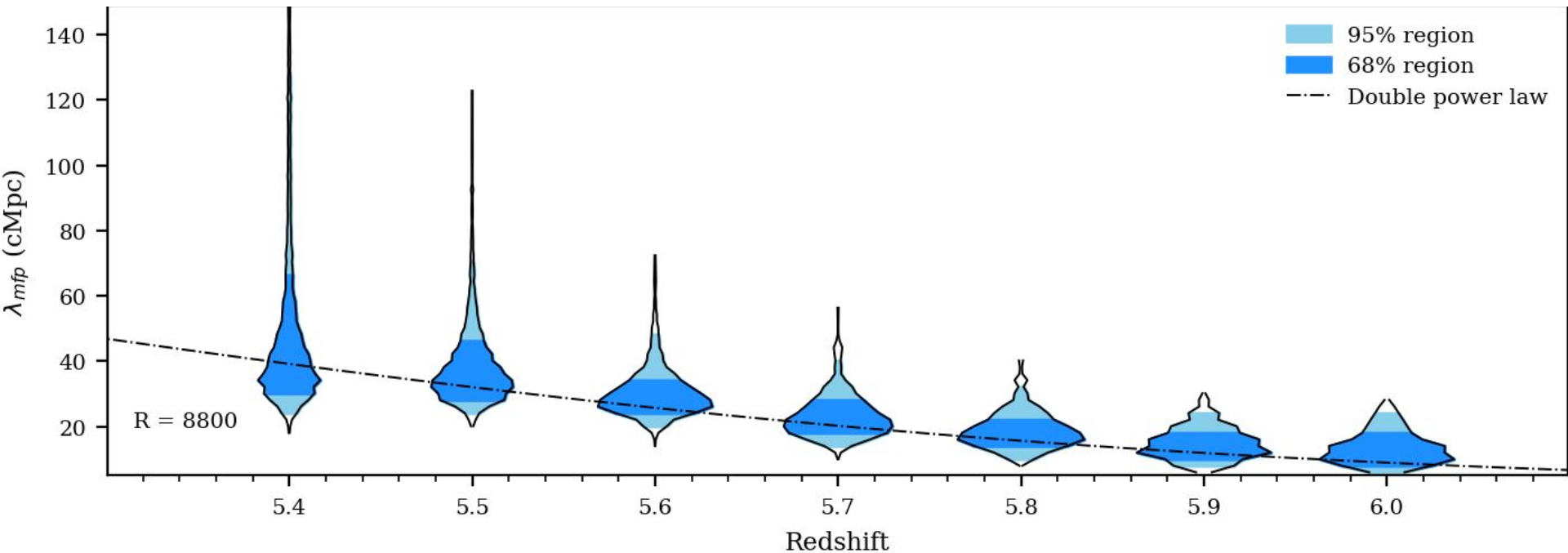
# Measuring $\lambda_{\text{mfp}}$ from mock data

Different redshifts have different precision due to:

- Size of the mock data
- The sensitivity of the autocorrelation function at the inferred  $\lambda_{\text{mfp}}$
- The luck of the draw



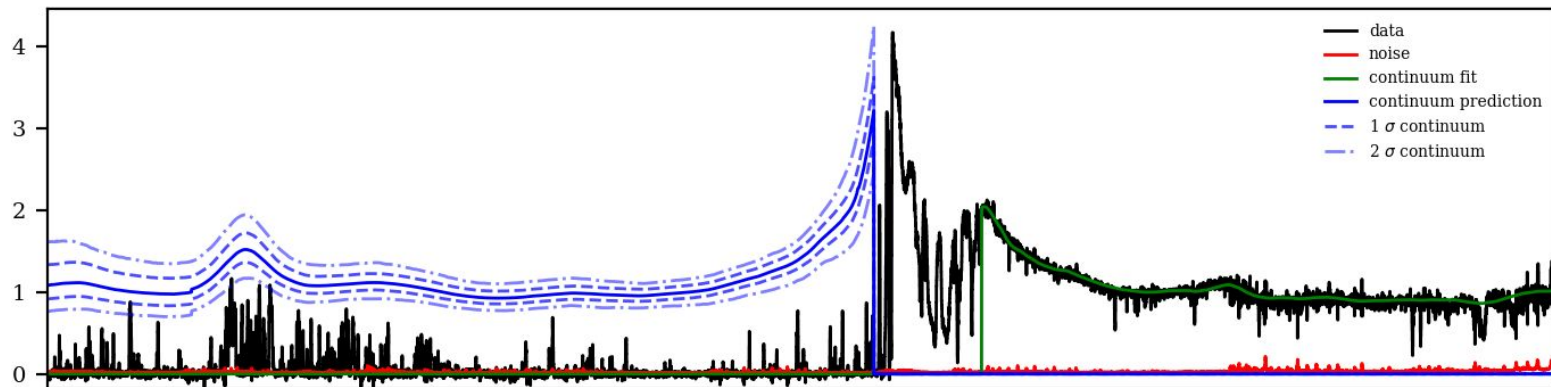
# Measuring $\lambda_{\text{mfp}}$ from mock data (removing “luck of the draw”)



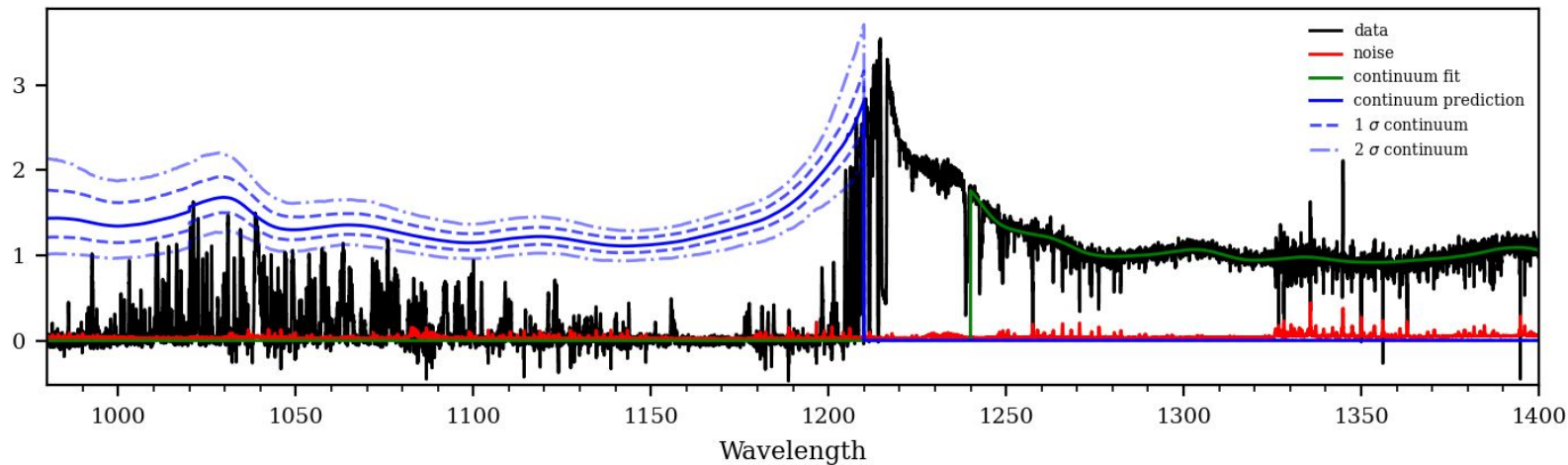
# Now with XQR-30 data:

PSOJ239-07,  $z = 6.1102$

Wolfson et al. in prep.



ATLASJ029.9915-36.5658,  $z = 6.021$



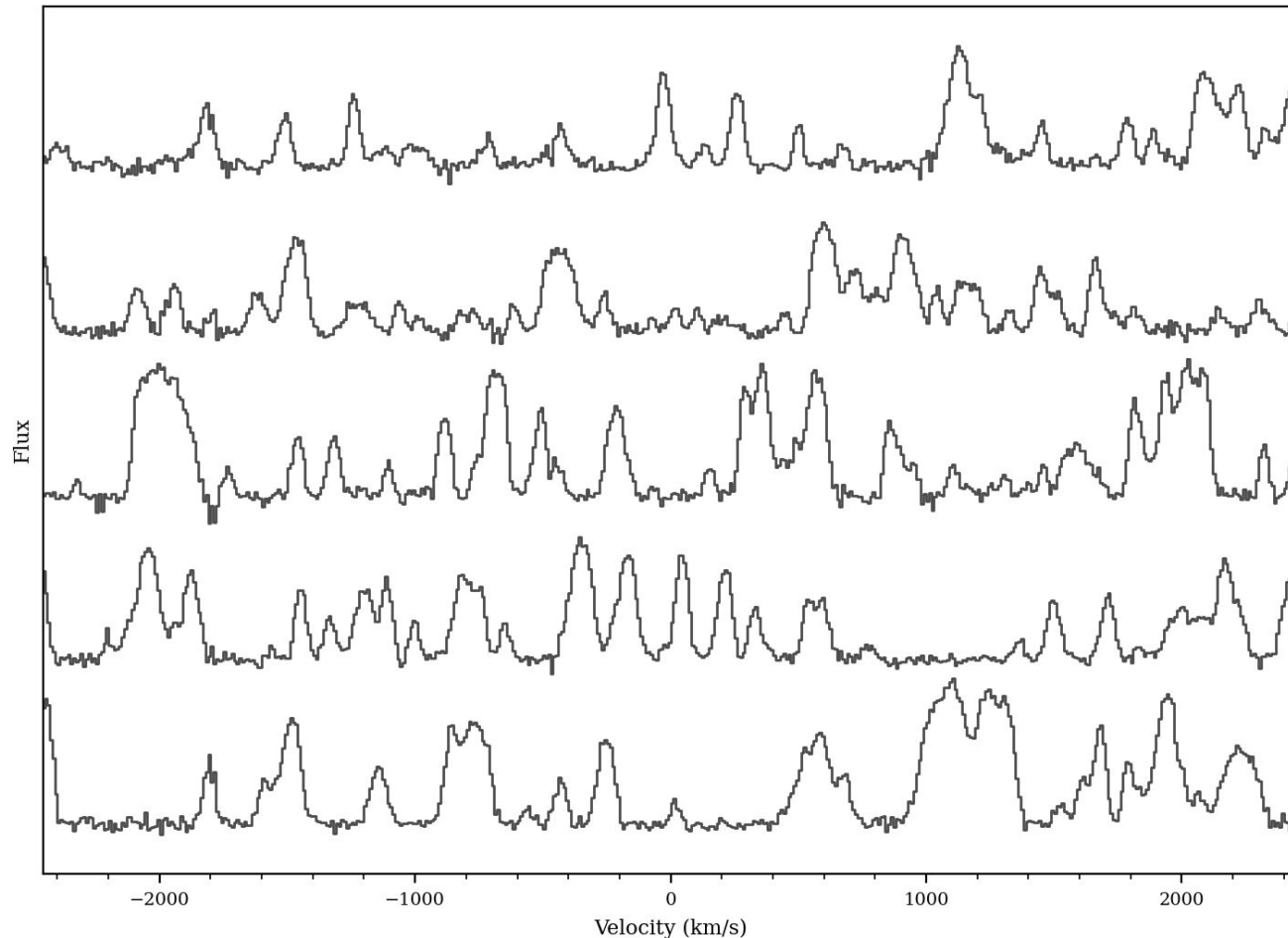


# Forward modeling XQR-30 data

Wolfson et al. in prep.

ATLASJ029.9915-36.5658

$z = 5.1, \Delta z = 0.1$

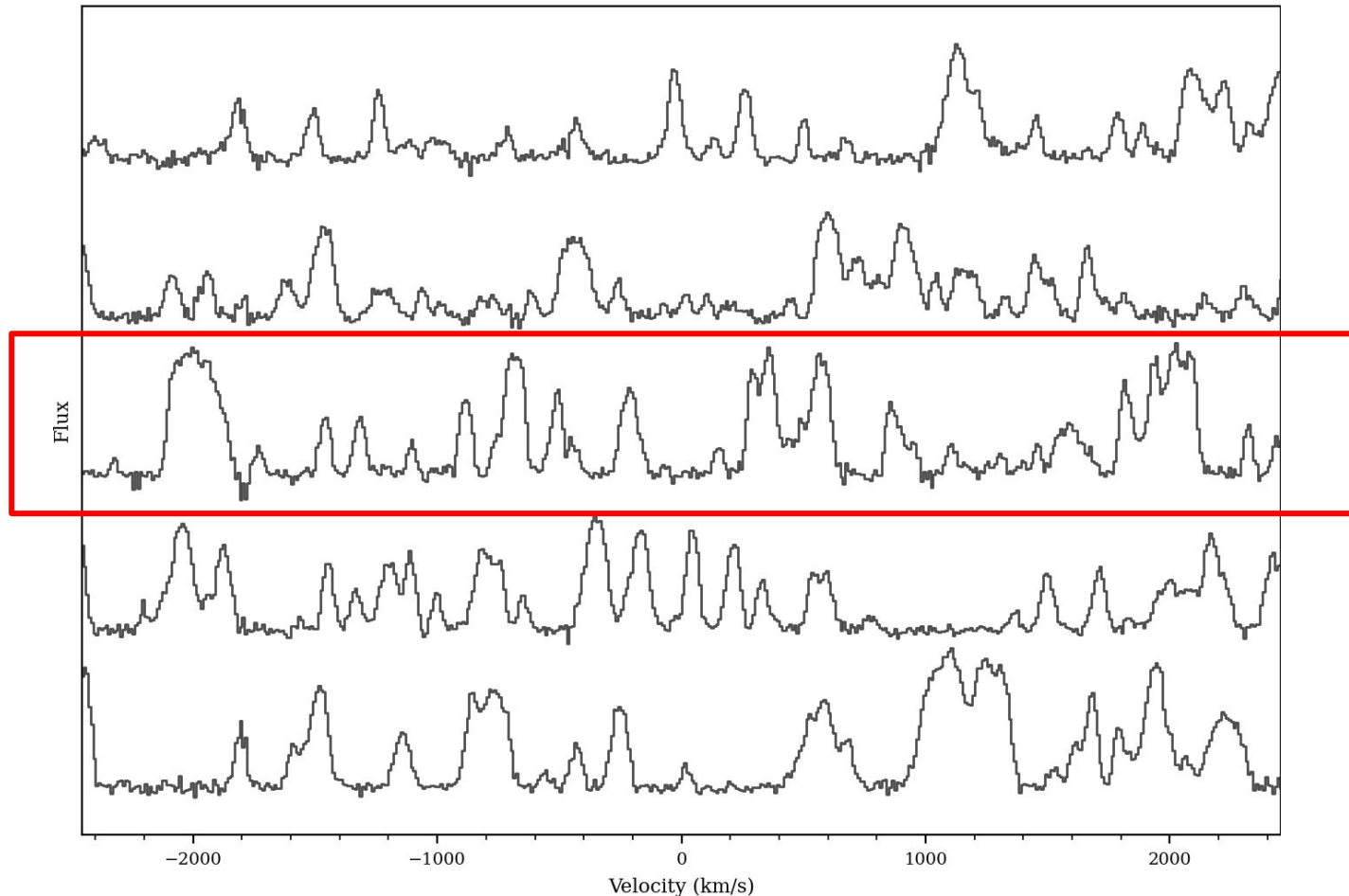


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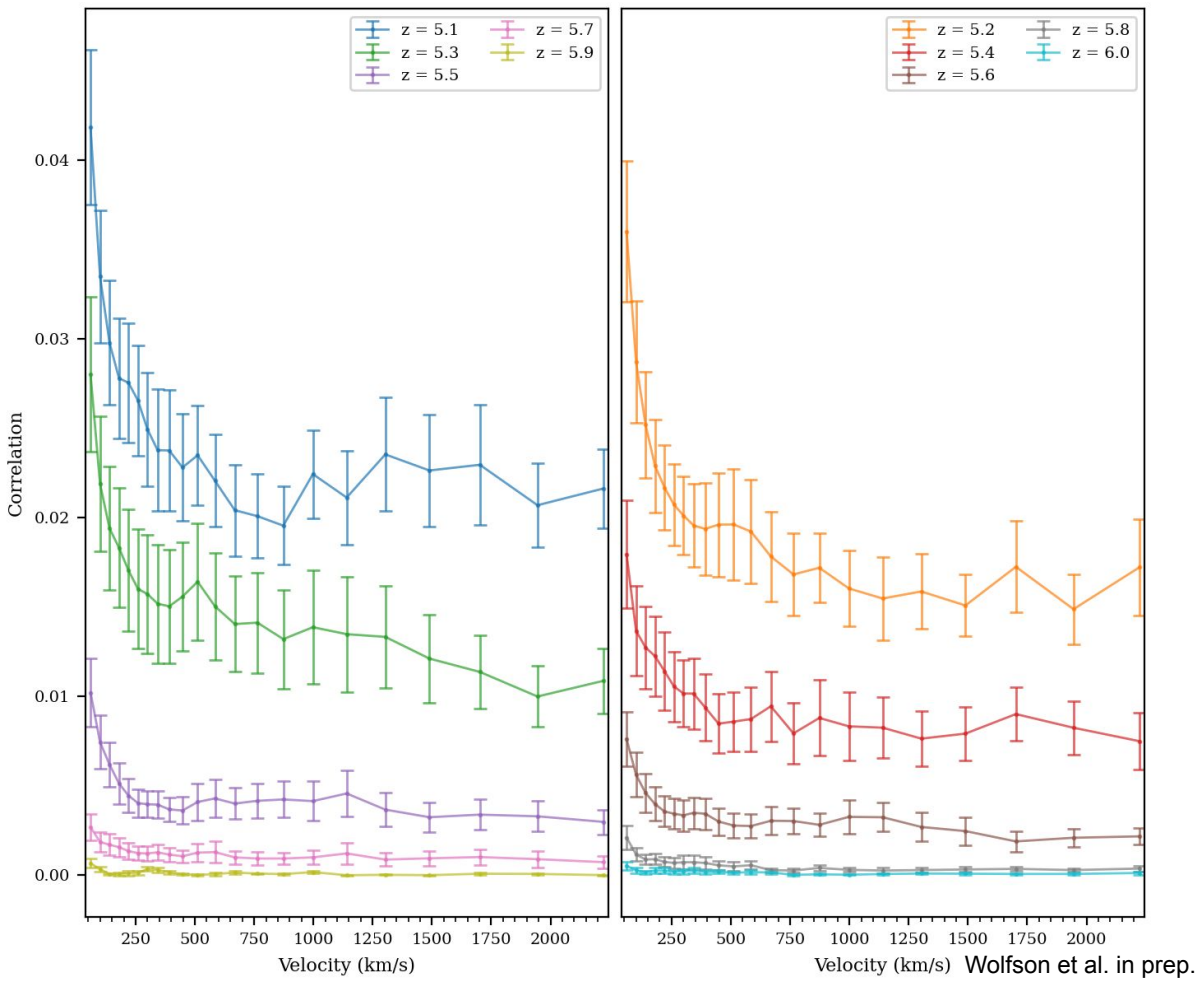
Wolfson et al. in prep.

ATLASJ029.9915-36.5658

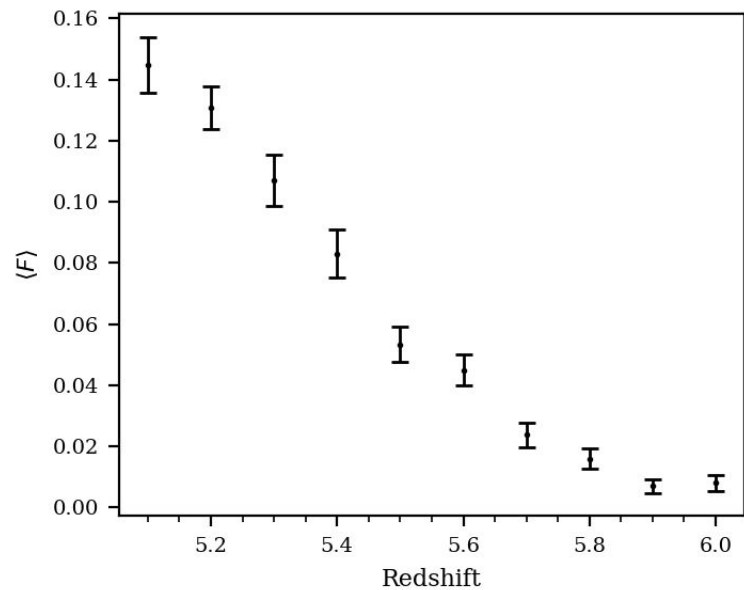
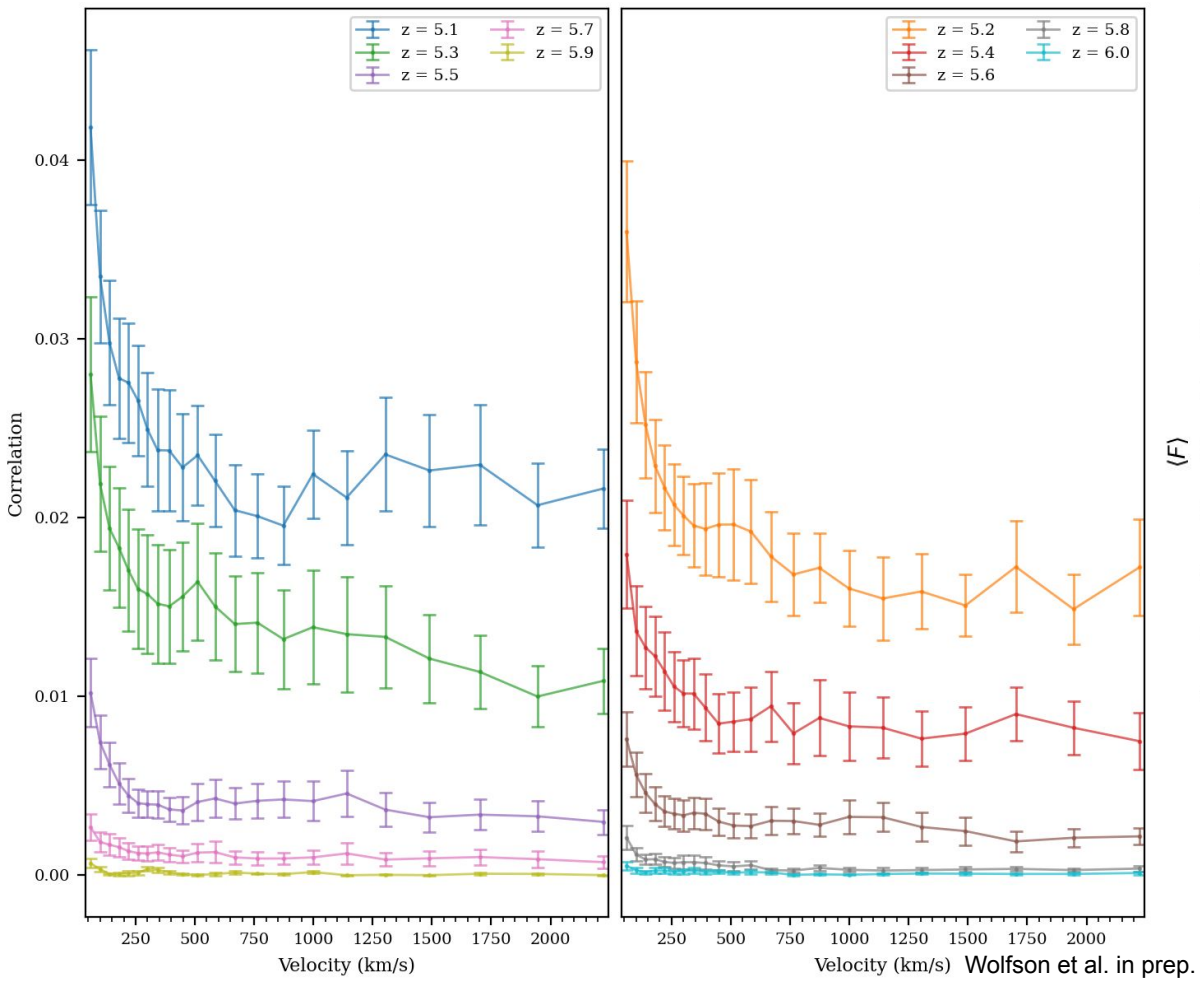
$z = 5.1, \Delta z = 0.1$



# Auto-correlation function from XQR-30

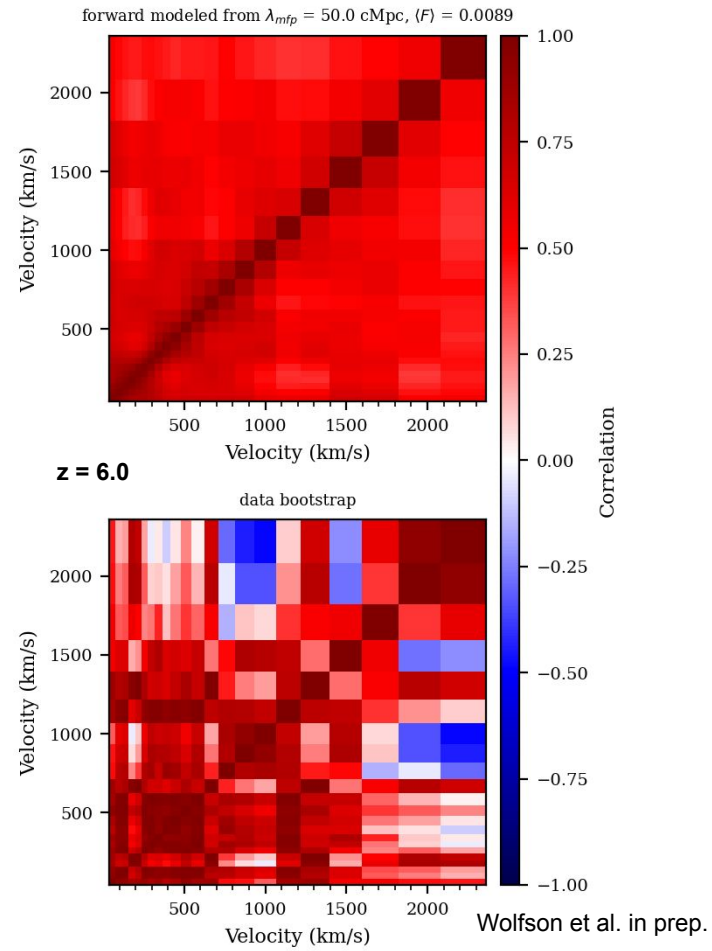
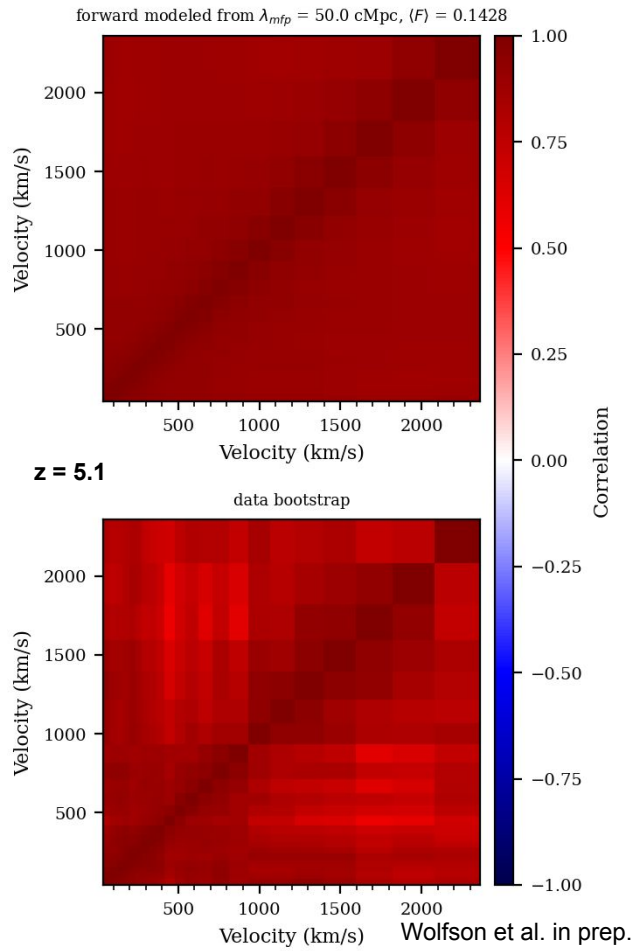


# Auto-correlation function from XQR-30



Wolfson et al. in prep.

# Calculating the covariance from noisy data is difficult

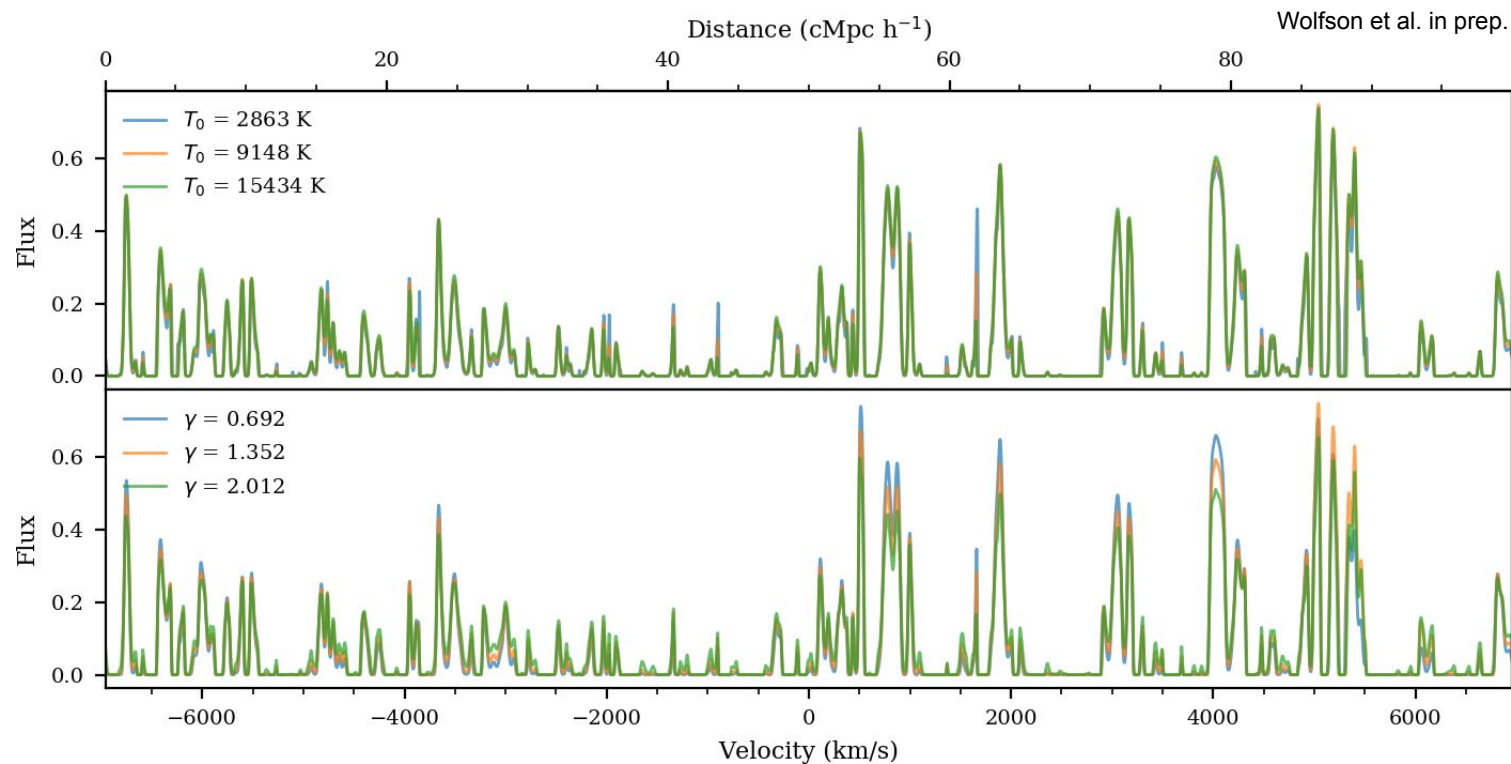




Is the auto-correlation function sensitive to the thermal state?

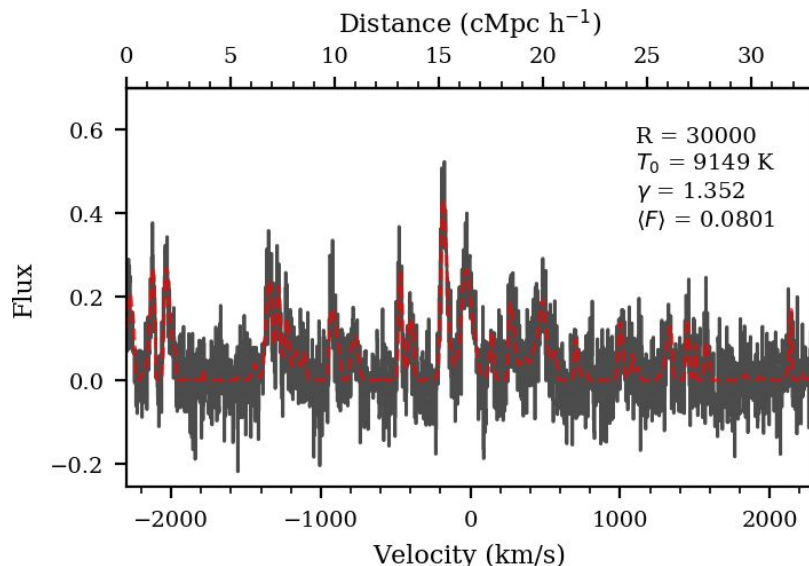
# Modeling the thermal state

- Same Nyx box with  $L_{\text{box}} = 100 \text{ cMpc h}^{-1}$
- Paint on different temperature-density relationships  $T = T_0(\rho/\bar{\rho})^{\gamma-1}$



# Modeling the thermal state

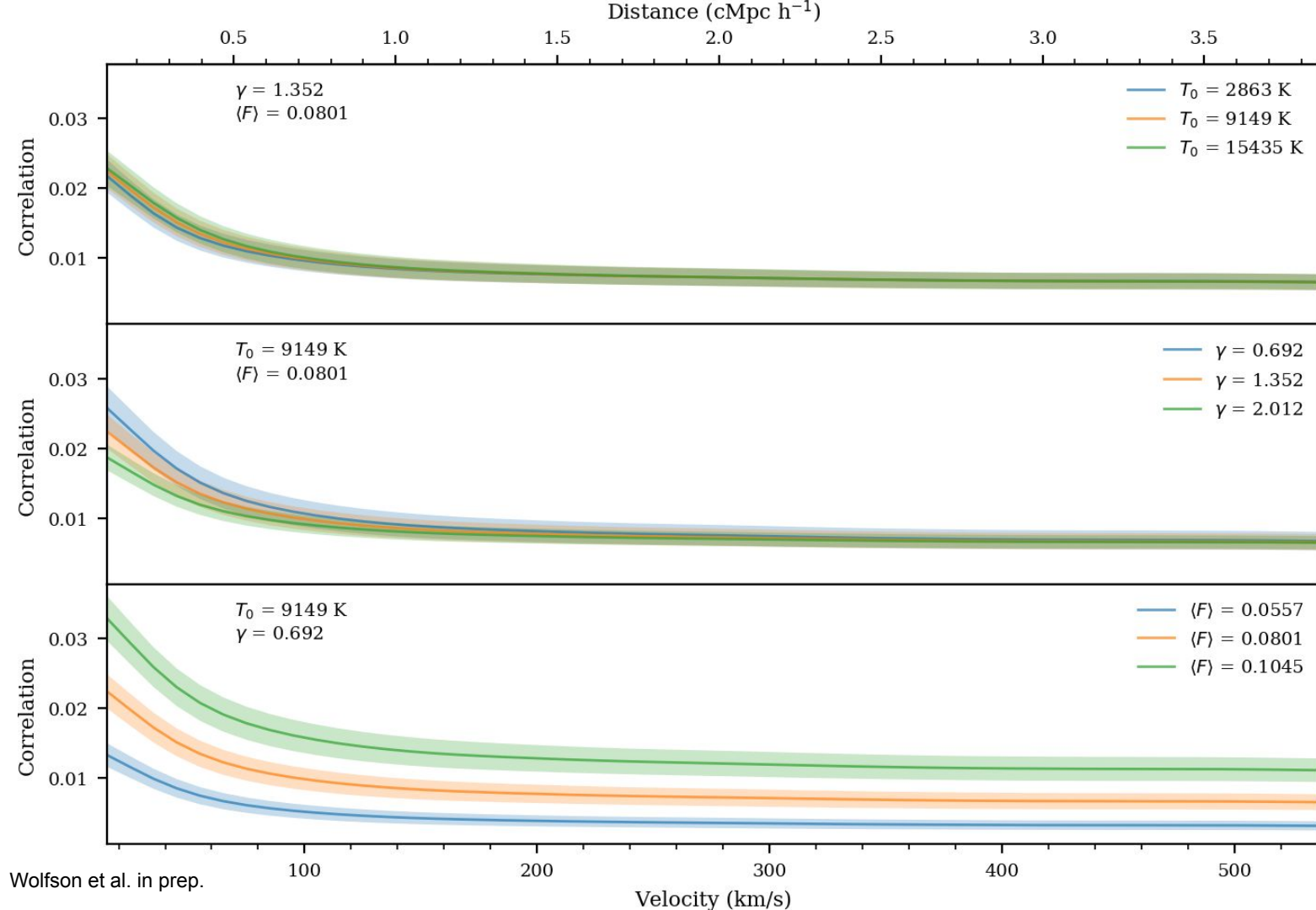
- Same Nyx box with  $L_{\text{box}} = 100 \text{ cMpc h}^{-1}$
- Paint on different temperature-density relationships  $T = T_0(\rho/\bar{\rho})^{\gamma-1}$



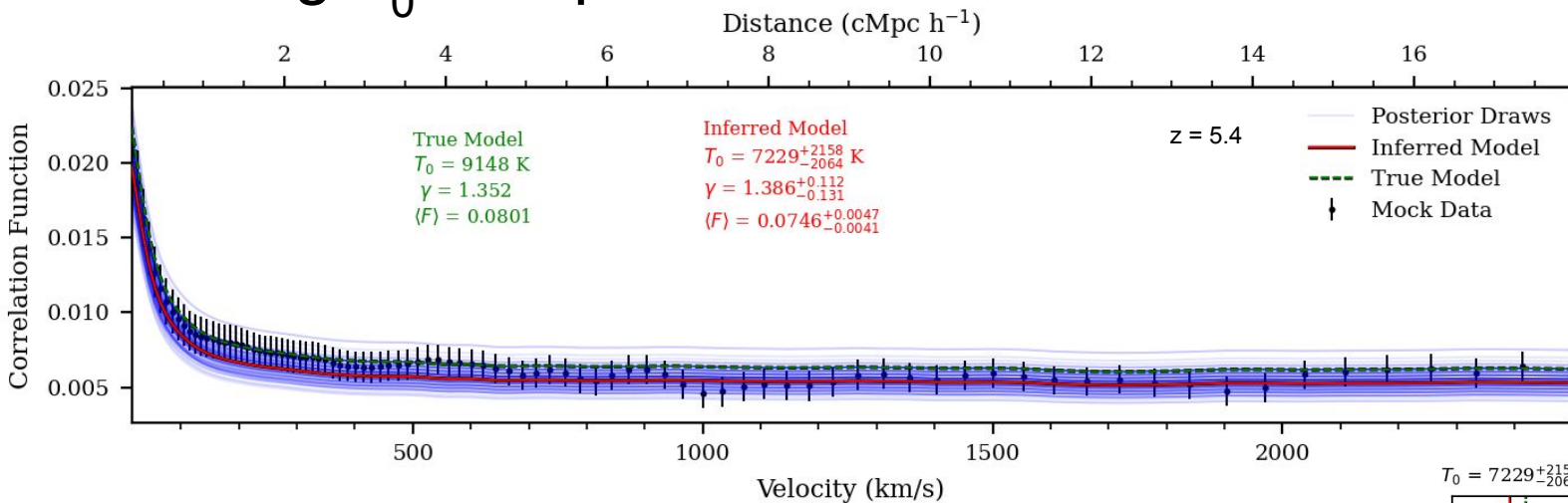
Wolfson et al. in prep.

## Mock data properties:

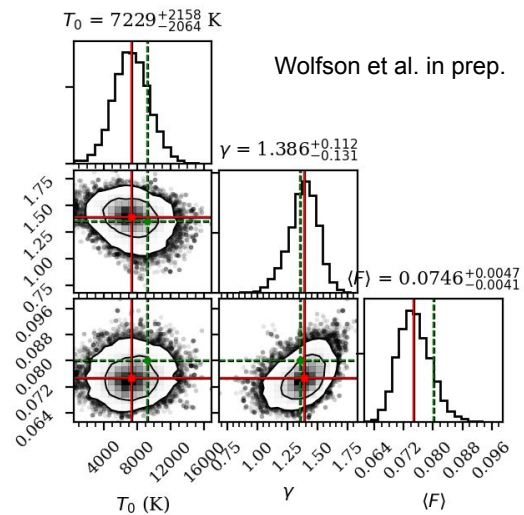
- High-resolution data  $R \sim 30000$ 
  - Resolution of Keck/HIRES
  - Convolve with a gaussian filter
- Gaussian noise with  $\text{SNR}_{\text{pixel}} = 10$
- 10 quasars in each  $z$  bin



# Measuring $T_0$ and $\gamma$ from mock data:

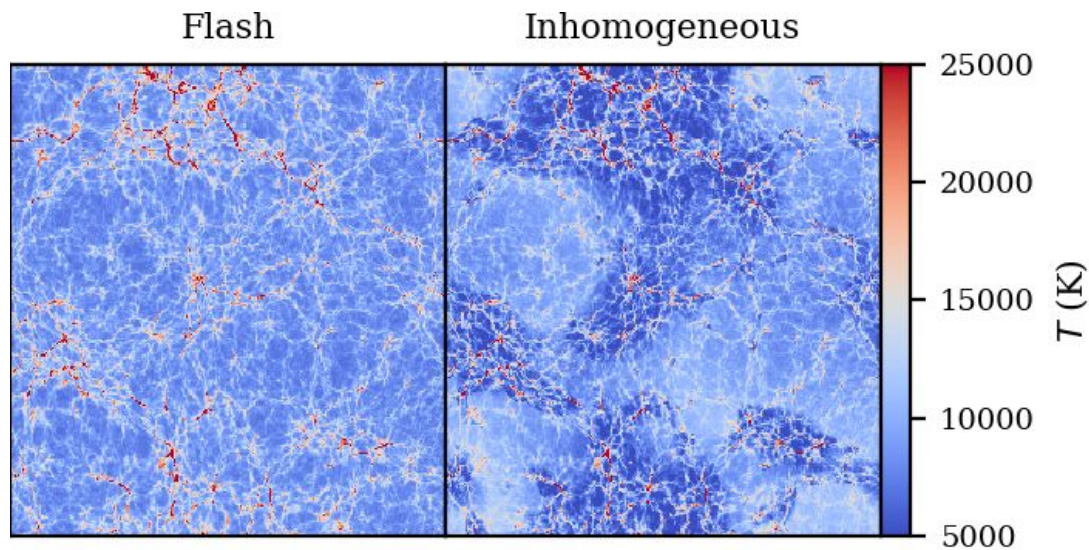


MCMC

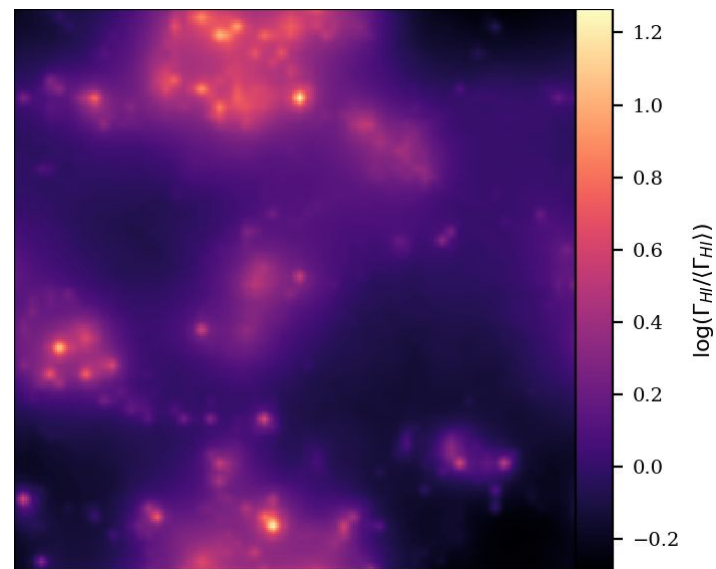




# Modeling different reionization scenarios



Wolfson et al. in prep.

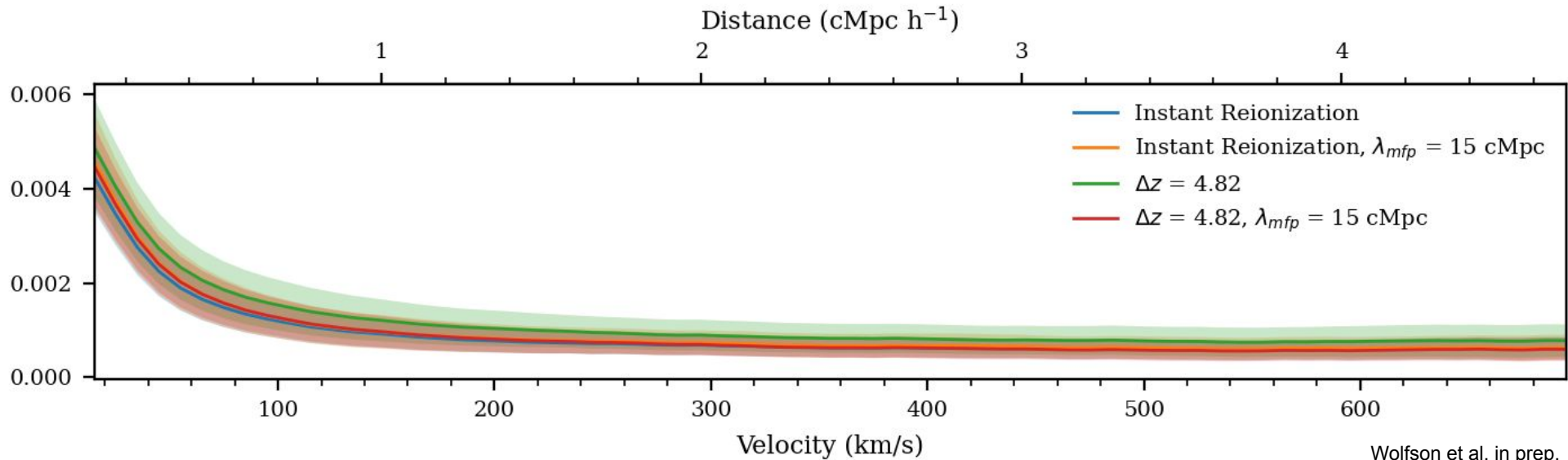


Wolfson et al. in prep.

Reionization models from Oñorbe et al. 2019

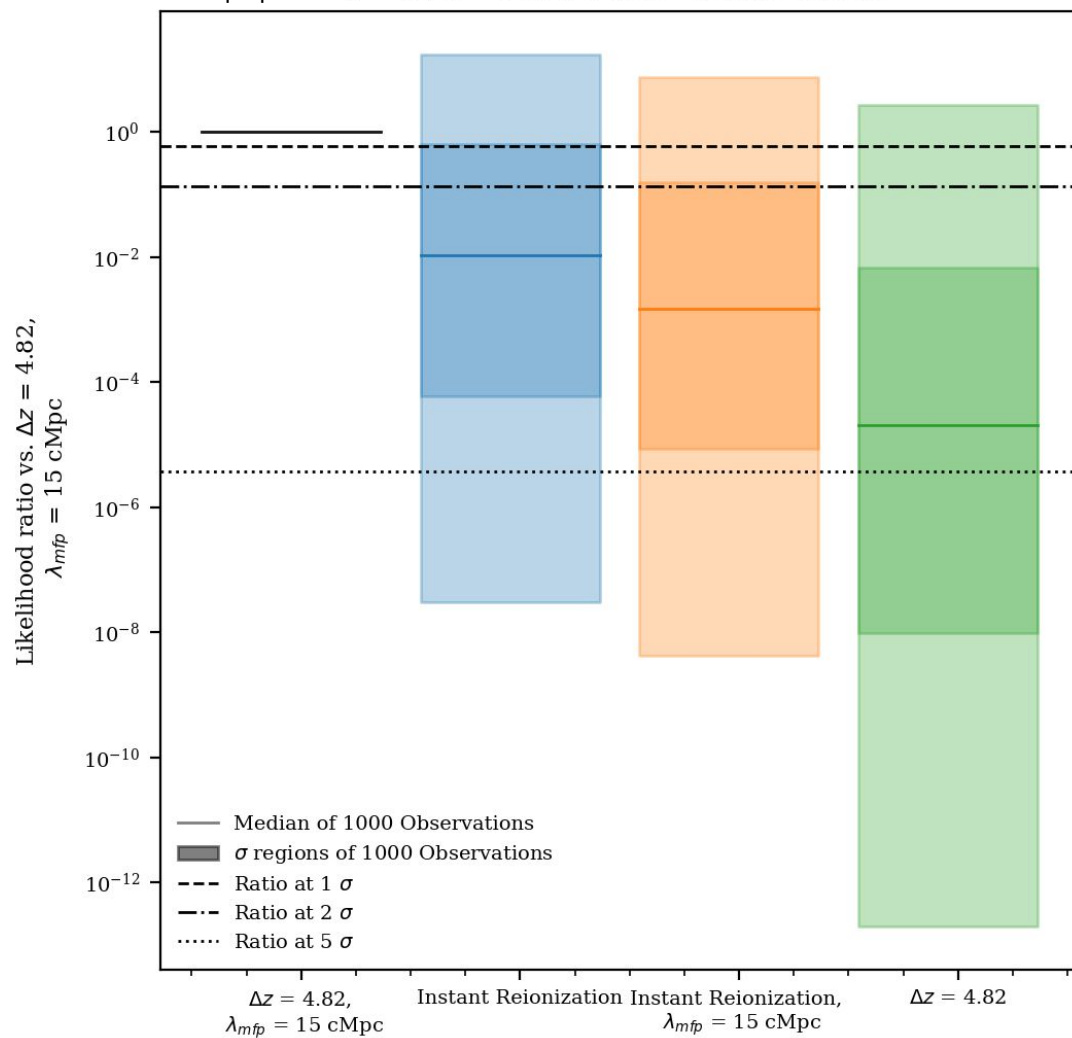
UVB box matching the IGM simulation with  $\lambda_{\text{mfp}} = 15 \text{ cMpc}$

# Auto-correlation from the four scenarios



# Likelihood ratio results

Can rule out other models  
at over  $2\sigma$  level for over  
50% of the mock data sets  
considered



# Summary

- The auto-correlation function provides a new way to competitively constrain the evolution with redshift of  $\lambda_{\text{mfp}}$  with existing data
- Practical considerations make the auto-correlation function a particularly useful statistic
  - Ex: not needing to model the noise or calculate the window function from DLA mask
- We can reweight the posteriors from mock data to correct for assumptions in our likelihood function (such as using a multivariate gaussian distribution)
  - However, further work on likelihood-free inference (LFI) should further improve precision of measurements
- Measurement of  $\lambda_{\text{mfp}}$  from XQR-30 data is ongoing
  - Estimating covariance matrix from limited data has been challenging
- The auto-correlation function will also provide a new way to constrain the thermal state of the IGM with high-resolution data

**Future work** – measure the auto-correlation function with high resolution data, WDM sensitivity, LFI